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FOR WIND INTEGRATION IN ERCOT***

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Acronyms and Abbreviations

AC	Alternating Current
AEP	American Electric Power
AMI	Advanced Meter(ing) Infrastructure
AMS	Advanced Meter(ing) System
API	Application Programming Interface
BESS	Battery Energy Storage System
BMS	Battery Management System
CCET	Center for the Commercialization of Electric Technologies
CPP	Critical Peak Pricing
CREZ	Competitive Renewable Energy Zone
CT	Current Transducer
DAQ	Data Acquisition
DC	Direct Current
DG	Distributed Generation
DNP3	Distributed Network Protocol 3
DOE	Department of Energy
DR	Demand Response
DSL	Digital Subscriber Line
DST	Discovery at Spring Trails
EID	Energy Internet Demonstration
EMS	Energy Management System
ePDC	Enhanced Phasor Data Concentrator
EPG	Electric Power Group
ERCOT	Electric Reliability Council of Texas
ETT	Electric Transmission of Texas
EV	Electric Vehicle
FERC	Federal Energy Regulatory Commission
FRRS	Fast Response Regulation Service
GNIRE	Group NIRE (National Institute for Renewable Energy)
GSEC	Golden Spread Electric Cooperative
HAN	Home Area Network
HEM	Home Energy Management
HVAC	Heating, Ventilation, and Air Conditioning
Hz	Hertz
ICCP	Inter-Control Center Communications Protocol
IEC	International Electrotechnical Commission

IEEE	Institute of Electrical and Electronics Engineers
IR	Interagency Report/ Internal Report
IRB	Internal Review Board
ISG	Intelligent Synchrophasor Gateway
ISO	Independent System Operator
kA	Kiloampere
kHz	Kilohertz
kV	Kilovolt
kVA	Kilovolt-ampere
kVAR	Kilovolt-ampere reactive
kW	Kilowatt
kWh	Kilowatt Hour
LMO	Lithium Manganese Oxide
LMP	Locational Marginal Pricing
LR	Legends Ranch
MHz	Megahertz
MVA	Megavolt Ampere
MW	Megawatt
MWh	Megawatt Hour
NEPA	National Environmental Policy Act
NIST	National Institute of Standards and Technology
NOAA	National Oceanic and Atmospheric Administration
PCT	Programmable Communicating Thermostat
PDC	Phasor Data Concentrator
PEV	Plug-in Electric Vehicle
PF	Power Factor
PGDA	Phasor Grid Dynamics Analyzer
PI	Principal Investigator
PM	Project Manager
PMO	Project Management Office
PMU	Phasor Measurement Unit
PSCAD	Power System Computer Aided Design
PV	Photovoltaic (Solar)
QSE	Qualified Scheduling Entity
REP	Retail Electric Provider
RLC	Resistive Inductive Capacitive
ROS	Reliability Operations Subcommittee
RTC	Reese Technology Center
RTDMS	Real-Time Dynamics Monitoring System
RTO	Regional Transmission Organization

RTU	Remote Terminal Unit
SEP	Smart Energy Profile
SGCT	Smart Grid Computational Tool
SF	Security Fabric
SFTP	Secure File Transfer Protocol
SGRDP	Smart Grid Regional Demonstration Project
SMT	Smart Meter Texas
SOAP	Simple Object Access Protocol
SPEC	South Plains Electric Cooperative
SPP	Southwest Power Pool
SSL	Secure Sockets Layer
SwRI	Southwest Research Institute
SWiFT	Scaled Wind Farm Technology
TBR	To Be Reviewed
TBD	To Be Determined
TDSP	Transmission and Distribution Service Provider
TDD	Total Demand Distortion
TDU	Transmission and Distribution Utility
THDv	Total Harmonic Distortion (voltage)
THDi	Total Harmonic Distortion (current)
TMS	Transmission Management System
TOU	Time Of Use
TPR	Technology Performance Report
TTU	Texas Tech University
VPN	Virtual Private Network
UCM	Universal Circuit Monitor
URD	Underground Residential Distribution
UTA	University of Texas at Arlington
WAN	Wide Area Network
WiSE	Wind and Science Engineering

EXECUTIVE SUMMARY

Introduction

Texas has for more than a decade led all other States in the U.S. with the most wind generation capacity on the U.S. electric grid. The State recognized the value that wind energy could provide, and committed early on to build out the transmission system necessary to move power from the windy regions in West Texas to the major population centers across the state. It also signaled support for renewables on the grid by adopting an aggressive renewable portfolio standard (RPS). The joining of these conditions with favorable Federal tax credits has driven the rapid growth in Texas wind capacity since its small beginning in 2000. In addition to the major transmission grid upgrades, there have been a number of technology and policy improvements that have kept the grid reliable while adding more and more intermittent wind generation. Technology advancements such as better wind forecasting and deployment of a nodal market system have improved the grid efficiency of wind. Successful large scale wind integration into the electric grid, however, continues to pose challenges. The continuing rapid growth in wind energy calls for a number of technology additions that will be needed to reliably accommodate an expected 65% increase in future wind resources.

The Center for the Commercialization of Electric Technologies (CCET) recognized this technology challenge in 2009 when it submitted an application for funding of a regional demonstration project under the Recovery Act program administered by the U.S. Department of Energy¹. Under that program the administration announced the largest energy grid modernization investment in U.S. history, making available some \$3.4 billion in grants to fund development of a broad range of technologies for a more efficient and reliable electric system, including the growth of renewable energy sources like wind and solar. At that time, Texas was (and still is) the nation's leader in the integration of wind into the grid, and was investing heavily in the infrastructure needed to increase the viability of this important resource. To help Texas and the rest of the nation address the challenges associated with the integration of large amounts of renewables, CCET seized on the federal opportunity to undertake a multi-faceted project aimed at demonstrating the viability of new "smart grid" technologies to facilitate larger amounts of wind energy through better system monitoring capabilities, enhanced operator visualization, and improved load management. In early 2010, CCET was awarded a \$27 million grant, half funded by the Department of Energy and half-funded by project participants. With this funding, CCET undertook the project named ***Discovery Across Texas*** and has demonstrated how existing and new technologies can

¹ In 2005, a consortium of Texas utilities, technology companies and universities formed CCET in Austin, Texas, to advance technology developments in the electric utility industry of Texas. During the next four years CCET completed a number of demonstration projects such as substation automation, demand response leveraging advanced metering systems, deployment of an initial synchrophasor system, and field testing of new remote terminal units. The seven demonstrations summarized in this report evolved from this experience and from leveraging new technology developments being deployed by member companies and the grid operator. That is, CCET leveraged other technology deployments in Houston, Dallas, Fort Worth, Austin and Lubbock that included solar communities, a wind technology testing center, synchrophasor monitoring and visualization systems, electric vehicles and smart homes.

better integrate wind power into the state's grid. The following pages summarize the results of seven technology demonstrations that will help Texas and the nation meet this wind integration challenge.

Wind Power in Texas

Statewide, wind power capacity in Texas has grown by more than 9,500 percent, from 116 megawatts (MW) in 2000 to 11,493 MW as of October 2014, with interconnection agreements in place for an additional 8,377 MW in 2015 as shown in the figure below.

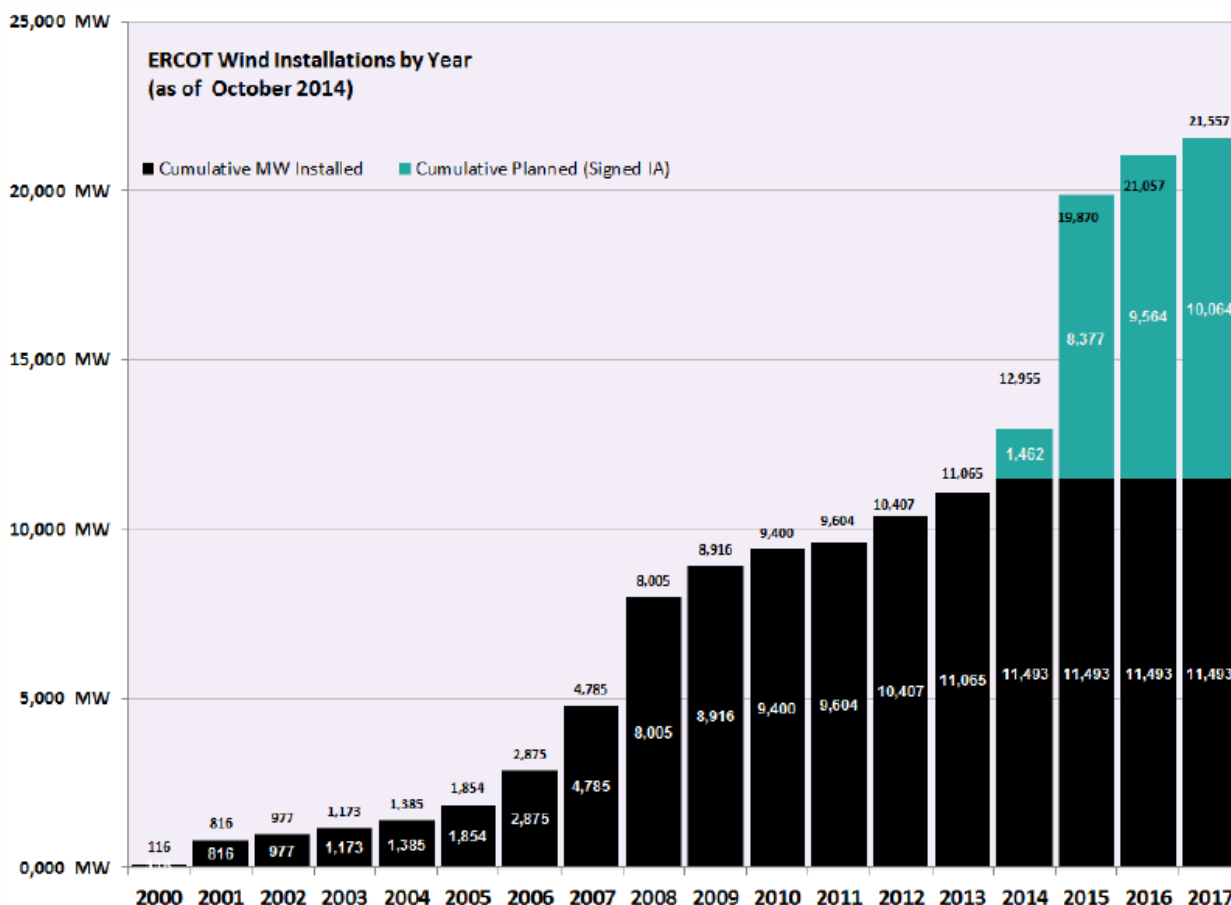


Figure 1. ERCOT Wind Capacity Installations

The installed wind capacity provides about 10% percent of the total grid generating capacity. In November of 2014, a new ERCOT wind record was set at 10,301 MW which represented nearly 33% of the load at that time. The increase in wind resources on the grid has required the development of about 3,600 miles of new transmission lines (shown in the figure below) into West Texas and the Panhandle area to move the power generated by those wind farms to the population centers in other areas of Texas. This five-year \$7 billion initiative was recently completed as part of an effort called Competitive Renewable Energy Zones (CREZ). Since the CREZ plan was developed to expand the transmission

capacity to accommodate 18.5 GW of wind farms, it may well have to be expanded to accommodate an even larger wind capacity since the State already has interconnection agreements in place for 21.5 GW of wind resources.

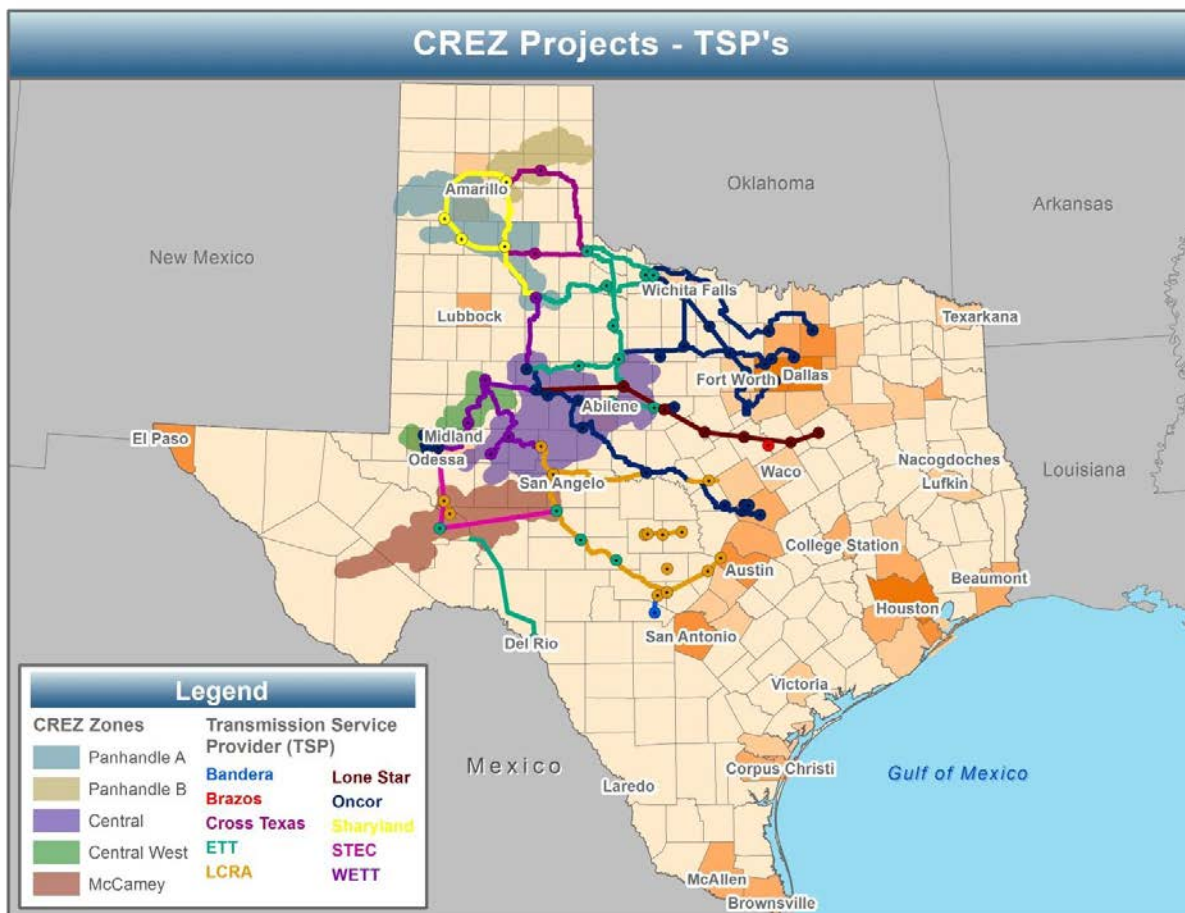


Figure 2. CREZ Transmission System Build Out

Project Location – Texas

This CCET project was primarily performed in an area encompassing about 85% of the state where the electric grid is managed by the Electric Reliability Council of Texas (ERCOT). The project also included deployment and testing of a utility-scale battery, some additional synchrophasor efforts, and cyber security protection systems in the Northwest area of Texas (Panhandle) that is actually in the Southwest Power Pool (SPP) region. These areas are shown in the figure below.

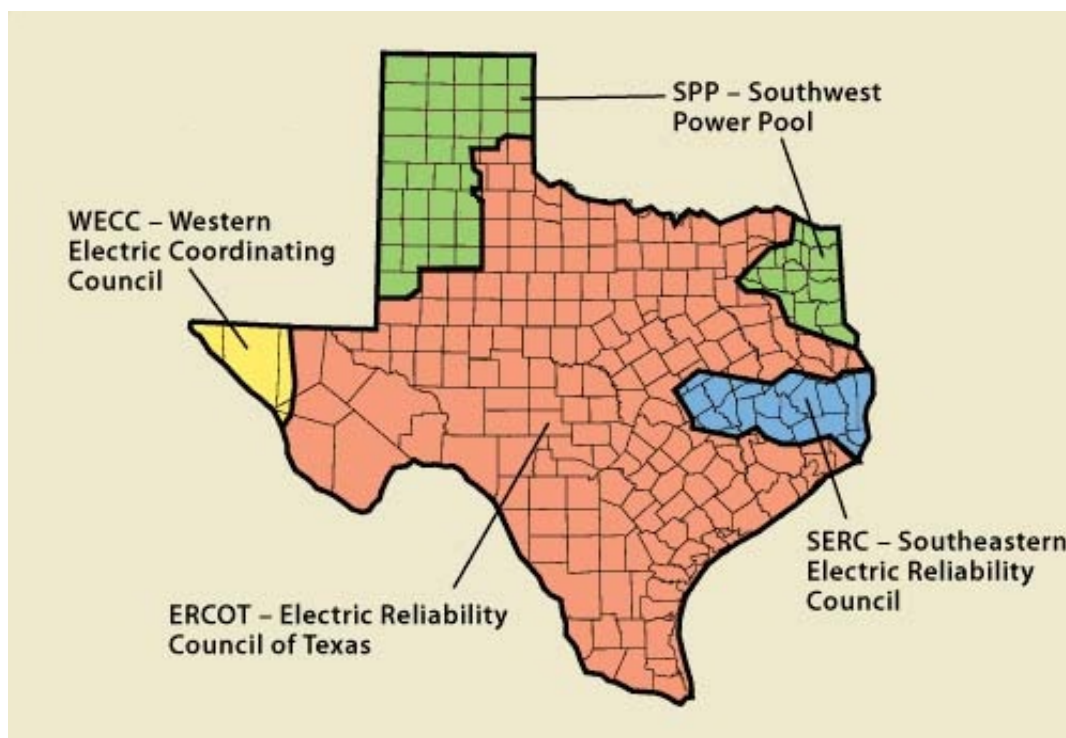


Figure 3. ERCOT and SPP Regions of Texas

Texas is unique among other regions with its high penetration of remote wind generation, one interconnection (ERCOT) entirely contained in the state, and one Independent System Operator (ISO), ERCOT. The ERCOT system is not interconnected across the state boundaries (except for three DC ties) and thus not subject to most regulatory rules of the Federal Energy Regulatory Commission (FERC). Texas deregulated the wholesale generation market in 1995 and opened the retail market to competition in 2002. As a result the generation and retail markets are now competitive while the transmission and distribution service providers (TDSPs) remain regulated by the Public Utility Commission. The investor-owned utilities were deregulated by the Texas Legislature while cooperatives and municipally-owned utilities were given the option to join the deregulated markets. ERCOT oversees and facilitates the competitive generation market where as many as 550 generation units compete for the provision of power on a daily basis and the retail markets where as many as 179 retail electric providers (REPs) compete for customer services.

In addition to leading the nation in wind generation plans, Texas also leads in the implementation of advanced metering systems. The most recent tally for the ERCOT region as of September 2014 indicates at least 6,849,921 smart meters have been deployed, and there are plans for installation of even more smart meters in future years.

Another unique aspect of Texas is the availability of the web-based Smart Meter Texas (SMT) Portal which hosts the 15-minute incremental meter data from many of those smart meters and is used for settlement purposes as well as permitting customers to view and share their current and historical

electrical usage information, and to bond home area network devices to their smart meter in support of demand response efforts.

Project Scope - Leverage Existing and Emerging Technology Resources

The Discovery Across Texas (DAT) project was carried out on real-time systems that leveraged a number of technologies already in place or under development in Texas. These technologies were:

- Initial synchrophasor network that was greatly expanded with more phasor measurement units, phasor data concentrators and enhanced visualization and monitoring software
- Wind testing facility with wind turbines at the Reese Technology Center in Lubbock
- Existing battery energy storage system (BESS) vendor with proven capabilities
- The CREZ transmission system expansion supporting wind farms throughout West Texas
- Emerging cyber security software from Intel/McAfee
- More than six million smart meters being installed throughout Texas
- Solar communities in Houston and Austin, the latter at the Austin Mueller Community which has rooftop solar, smart appliances and electric vehicles
- Existing programmable thermostats and gateways, and available consumer participants, involved in demand response programs in Dallas and Houston
- Smart Meter Texas (SMT) portal for hosting and displaying consumer smart meter data
- Fleet of electric trucks owned and operated by Frito Lay in Fort Worth

These technologies facilitated the implementation and execution of seven different project components, or technical demonstration areas:

- Synchrophasor monitoring, visualization and event reporting
- Cyber security protection for synchrophasor data streams (Security Fabric)
- Distribution-level battery energy storage system
- Residential time-of-use pricing trials
- Circuit and transformer monitoring of residential solar
- Residential demand response using direct load control devices
- Fast response regulation service with fleet electric vehicles

The efforts and accomplishments of these project demonstrations are highlighted below.

Improving Grid Reliability with Synchrophasors

As a renewable and clean source of energy, sustained winds can be harnessed using wind turbines to provide great amounts of power, but this source is intermittent with limited predictability, and there can be periods of both too much and too little wind. Through this project, CCET demonstrated that new monitoring technologies can help integrate more wind resources while more effectively managing wind

power's effects on the grid. To help determine power quality conditions and identify developing disturbances on the grid, CCET's demonstration project used synchrophasor technology which measures power variations in real time. Phasor measurement units, or PMUs, measure the grid's voltage and phase angles to provide utilities with near real-time views of grid conditions. By time-stamping each measurement, synchrophasors enable utilities and ERCOT to compare measurements from different times, as well as different locations, thereby tracking power flows in real-time, which can be useful for preventing disturbances and disruptions, including blackouts.

At the inception of this project, the ERCOT network was being monitored by PMUs at three locations. Synchrophasor data was being captured and streamed from those devices to a phasor data concentrator (PDC) at ERCOT. The data was then synchronized based on timestamps and displayed via a Real Time Dynamics Monitoring System (RTDMS), provided by Electric Power Group (EPG), which was running in the test environment at ERCOT. As part of this demonstration project, three utilities (Oncor Electric Delivery, American Electric Power, and Sharyland Utilities) committed to install PMUs at 13 additional locations, and to stream their synchrophasor data to ERCOT for monitoring and display in the RTDMS. After recognizing the true value proposition of this early warning network of devices, the utilities began adding PMUs at even more locations on their networks. As of November 2014, as shown in the figure below, the utilities had installed and are providing streaming synchrophasor data from 76 PMUs at 35 locations to the ERCOT RTDMS. These efforts also attracted the attention of other utilities across Texas who began planning for deployment of additional PMUs. Currently, ERCOT is anticipating receiving data from nearly 90 PMUs at 46 different locations by early 2015. This network of PMUs currently provides monitoring of about 78% of ERCOT's regional grid footprint, and this should increase to about 85% coverage by the end of the project.

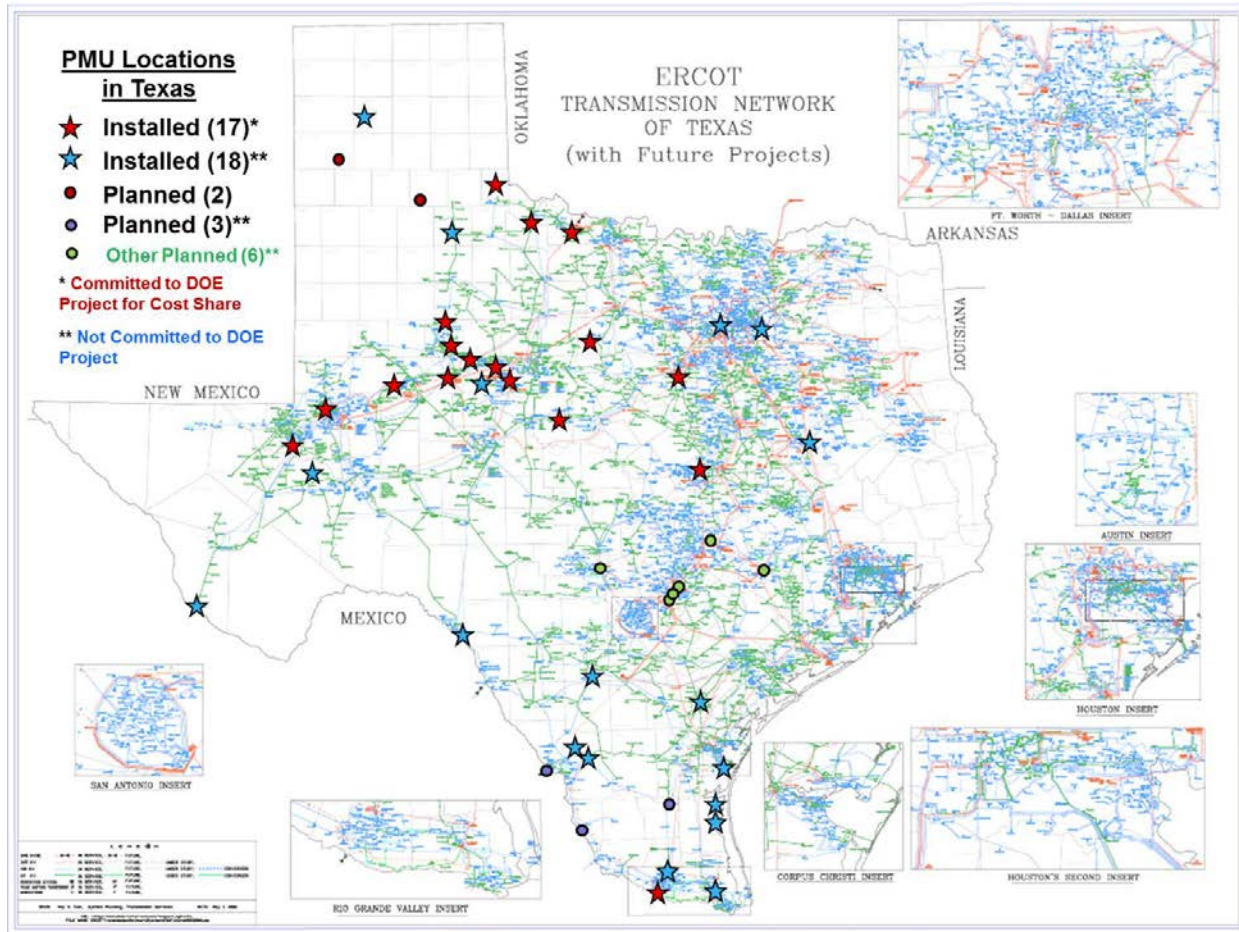


Figure 4. PMU Deployment in Texas

In addition to the ERCOT synchrophasor network, another objective of the project was to acquire data on the use of PMUs in tandem with wind turbines, and to evaluate their benefit to distribution grid operations. For this aspect of the project, Texas Tech University (TTU) worked with several electric cooperatives to deploy a synchrophasor network in the Texas Panhandle portion of SPP. Currently, this system consists of five PMUs, a PDC located at the Reese Technology Center and an RTDMS server located on the TTU campus. Two of these PMUs are also functioning as revenue meters for two wind generation assets: a 1.67 MW Alstom wind turbine and three 300 kilowatt (kW) wind turbines at the Sandia National Laboratories' Scaled Wind Farm Test (SWiFT) facility, while the third PMU is functioning as a revenue meter for a battery energy storage system (BESS) which is located near the SWiFT facility. The red stars in the figure below indicate the existing PMU locations, and the yellow stars indicate possible future locations, three of which are already planned and will significantly expand the geographical area of the network and provide PMUs located near significant wind or natural gas generation assets.

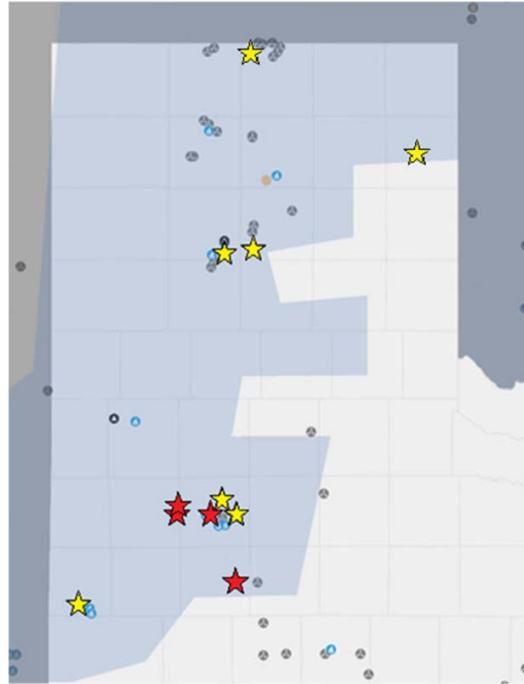


Figure 5. Current and Planned TTU PMU Locations

In addition to greatly expanding the synchrophasor network, a key goal of the project was to promote the benefits and prove that it could be beneficial to grid operators in the ERCOT control room. Through the collaborative efforts of the project team members, coupled with active participation by ERCOT and Texas utilities, Texas has become one of the first in the nation to use synchrophasor technology as an operator tool to improve grid reliability. The figure below shows the RTDMS in use at ERCOT.



Figure 6. RTDMS Provides Grid Reliability Indicators at ERCOT

By quickly pinpointing potential grid issues, synchrophasor technology based on readings 30 times per second helps utilities and ERCOT respond and compensate more quickly, thereby improving the reliability of wind energy and the grid as a whole. For example, in the figures below, the synchrophasor system quickly identified a potentially damaging grid oscillation originating from a malfunction at a wind farm in January 2014. This was an observation/detection not available from the traditional, slower SCADA systems. The unit was constrained by the ERCOT grid operators who stopped the oscillation, and upon notice the wind farm operator quickly corrected the malfunction. This use of the RTDMS demonstrated that, during and after an energy oscillation, the grid operators can use synchrophasor data to trace a problem to its source, and typically identify the cause of the issue.

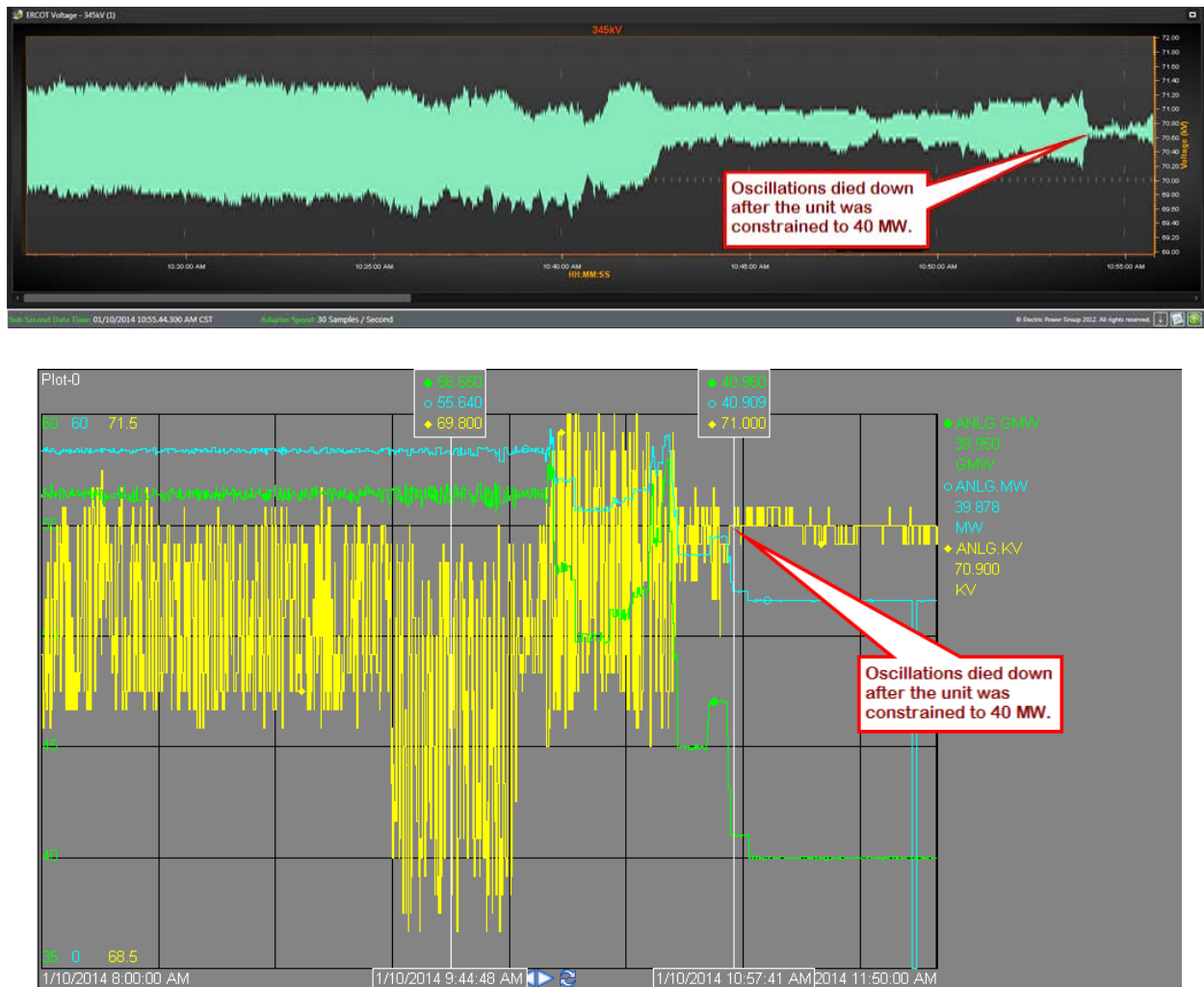


Figure 7. Reduction of Oscillations after Constraining the Plant to 40 MW on 10 January 2014

In addition to deploying large numbers of PMUs, and streaming that data to ERCOT for grid monitoring, the CCET project performed a number of studies and analyses to improve operations. These included:

- Baselining studies
- Data quality studies
- Lessons learned from mitigating PMU data loss
- Event analyses of generation outages and wind interactions
- Requirements for control room and network production upgrades
- ERCOT and utility operator training
- Slow scan synchrophasors
- Synchrophasor use cases

Another area of interest was in using synchrophasor data to improve generator model parameters. These mathematical models of generators are fundamental management tools for grid operators but sometimes the accuracy of such models are less than is needed for understanding the generator performance. The DAT project developed a computer algorithm that uses phasor data from near the generator to improve the parameter estimates of the generator model, and thus the accuracy of generation forecasts by ERCOT. Such computer-based automation capabilities are destined to become standard tools for generator owners and grid operators of the future.

ERCOT presently utilizes synchrophasor data for system oscillation detection, generator model validation, and post event analysis and reporting. In the future, intended uses include voltage stability monitoring and other applications of synchrophasor data for grid operations.

Security Fabric for Protection of Synchrophasor and Other Grid Systems

Synchrophasor systems are destined to become control room tools for grid management and therefore will need to be protected from cyber attacks under a Presidential directive known as Critical Infrastructure Protection (CIP). Based on this need, the CCET project included an effort to incorporate and validate the effectiveness of enhanced security protection mechanisms in a synchrophasor network. For this effort, Intel/McAfee provided Security Fabric components that are architected to address the seven security tenets of the NIST-IR 7628. One of the keys of their approach was to employ a hardware/firmware solution that provides exceptional security for applications, and does not require that the applications be modified.

The project employed contemporary concepts of silicon-driven security to stop attacks that easily defeat software-only security attempts. This provides a more secure framework for incorporating security in the physical hardware layers to better protect sensitive operations i.e. providing authentication for every step of the boot sequence to eliminate infection by malware or root-level viruses. The project approach leveraged the hardware security, and provided whitelist attestation to verify the trust basis for all changes and transition control to promote changes at a safe and convenient time. The benefits of this approach include the following:

- Application whitelisting on management systems
- Movement of databases to secured management system with encrypted tunnel
- C37.118 data (standard for PMU accuracy and data communications) is only received from authorized nodes
- C37.118 data is exchanged between nodes using a secure tunnel
- Access to applications is limited to authorized clients
- Secure network time protocol communications

For this demonstration, EPG designed, integrated, validated, and tested the ability of the Security Fabric components to provide the required cyber security protection for its ePDC and RTDMS applications in a laboratory environment. Once the testing was completed, the Security Fabric servers together with the EPG software were deployed at Texas Tech University. The TTU had a similar configuration in place without the Security Fabric protections, so this new set of servers and EPG software was intended to provide a parallel environment, as shown below, that leveraged the real-time PMU data streams at TTU.

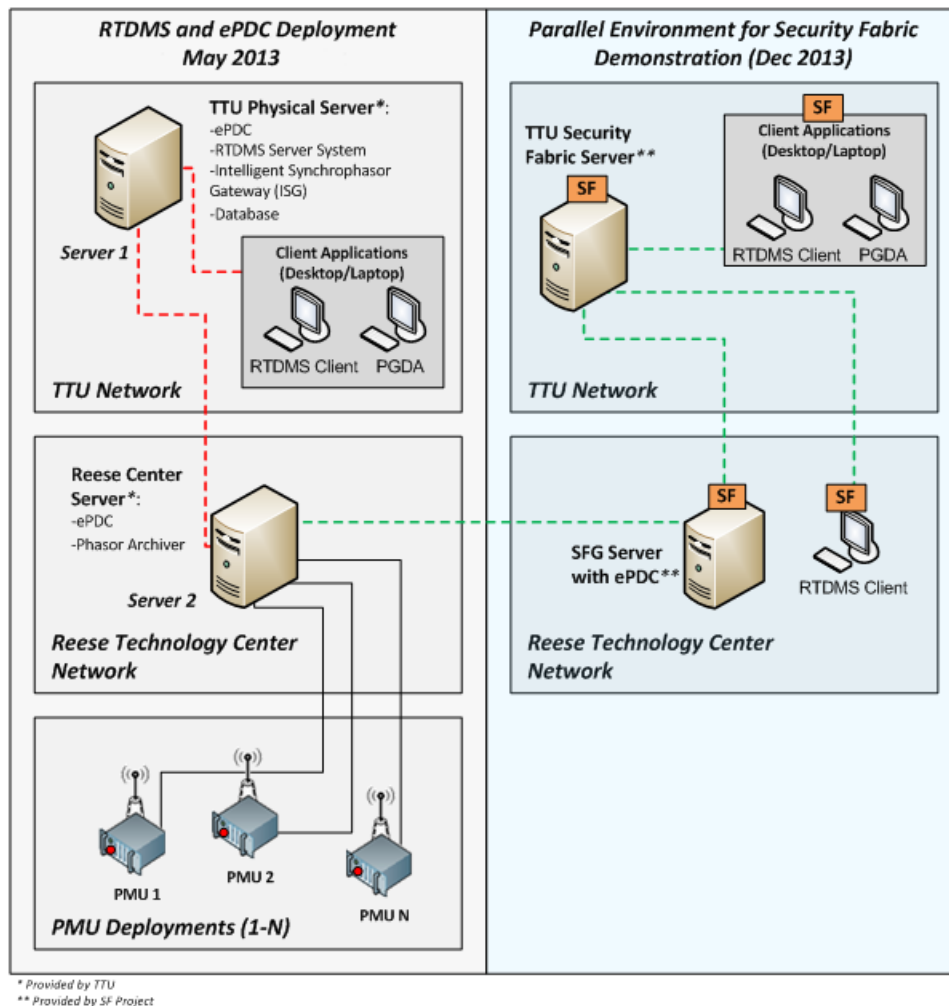


Figure 8. Security Fabric Test System at Texas Tech University and the Reese Technology Center

A field demonstration was performed on the TTU synchrophasor network to demonstrate the use of the Security Fabric to secure the phasor data stream between a “virtual Transmission Owner/Utility” and a “virtual Independent System Operator”. The goal of the demonstration was to validate that the phasor data stream could be successfully secured from cyber attack while still delivering timely phasor data to the “virtual ISO’s” RTDMS system. As a part of the demonstration, TTU faculty and students performed the following activities:

- The first activity was ***verification of security fabric requirement specifications***. This effort involved testing to determine whether the Security Fabric-enabled network at TTU satisfied the security requirement specifications provided by Intel/McAfee. The specifications were based on the “Guidelines for Smart Grid Cyber Security” NISTIR-7628 requirements.
- The second activity was ***penetration testing*** on the synchrophasor network without Security Fabric and the Security Fabric-enabled network. This involved coding scripts to identify security vulnerabilities and injecting attacks by exploiting selected vulnerabilities found.

For the first activity, Intel/McAfee provided Security Fabric relevant security specifications that cover 11 categories of the NISTIR 7628 security requirements. The Security Fabric-enabled network at TTU was tested against these specifications. The process of testing in this part involved three basic activities: (1) *requirement acquisition* (communicating and assisting in the interpretation of the testing requirements according to NIST standards), (2) *specification description* (providing feedback and clarification in defining proper security specifications), and (3) *specification verification* (verifying that the system satisfies the given security specifications). This process was time consuming due to the nature of evolving requirements and specifications development.

The results of this compliance testing, based on the details provided in the requirements specification, verified that 86.5% of the security specifications were satisfied. Additional testing on the part of TTU determined that the remaining requirements could be satisfied by 1) clarifying the specification so that the test intent, objectives, steps and results were adequately defined, and 2) incorporating additional existing Intel/McAfee tools or applications.

In the next activity, TTU performed penetration testing on both the Security Fabric-enabled network and the unsecured network. Interestingly enough, the total number of vulnerabilities of the Security Fabric-enabled network was 13 which was seven more than those found in the non-Security Fabric environment. This higher count is because the secure system includes additional Intel/McAfee software monitoring and management components that are also vulnerable. In the findings, TTU pointed out the following:

- All but one of the vulnerabilities identified in the Security Fabric-enabled network were caused by misconfiguration errors, or because of older, less secure versions of software, so these were easily fixed.
- The connection between the two synchrophasor endpoints, PDC and RTDMS, which was the real intent of this effort, proved to be more secure with Security Fabric.

Based on the findings of this effort, Intel/McAfee was able to resolve all of the identified vulnerabilities, and they have since updated the support guides to ensure that these potential vulnerabilities are mitigated during installation of the Security Fabric components.

Battery System Deployment on a Distribution System

Another important technology for wind integration is battery storage. For this effort, CCET was interested in assessing how battery storage can supplement wind turbines to improve wind-power efficiency. The overall objective was to test the ability of a battery energy storage system (BESS) to perform various needed grid functions, such as peak shaving, load leveling, and frequency support. A secondary objective was to determine the economic viability of deploying these systems to support grid operations. The Reese Technology Center near Lubbock, Texas, afforded an optimal location for satisfying the objectives since it was already established and outfitted as a wind test facility, it provided access to a distribution grid operated by a local cooperative that was very supportive of testing, and was close to TTU which had a staff capable of performing the BESS evaluations. The Team defined a detailed functional specification for a BESS, and then issued a competitive request for proposal. After negotiations, CCET awarded a contract to Xtreme Power (now Yunicos) and Samsung SDI for a 1MW Lithium Manganese Oxide (LMO) BESS. This system, pictured below, was built, tested, fielded, and connected to the grid in eight months, becoming fully operational in August 2013.



Figure 9. BESS Installed at the Reese Technology Center

The BESS houses 18 racks with 256 Lithium Manganese Oxide batteries totaling 4,608 battery cells with a combined energy storage capacity of 1 MWh.



Figure 10. Racks of Batteries within the BESS Container

The BESS operates in conjunction with about 4.6 MW of wind generation at the site. This includes an Alstom ECO86 1.67 MW wind turbine, a Gamesa 2 MW wind turbine, and three V27 300kW Vestas turbines that were deployed as part of the SWiFT facility which currently produces 900 kW, and will eventually be expanded to produce 3.6 MW. The site also includes a 200-meter meteorological tower to provide unique atmospheric measurements for wind analysis. Two images of the SWiFT facility are shown below. The first figure below shows a view of the three SWiFT turbines and the meteorological tower. The second figure shows an aerial view of the SWiFT turbines and the close proximity of the BESS.



Figure 11. SWiFT Facility at Reese Technology Center



Figure 12. Aerial View of the BESS and the SWiFT Facility Turbines

The BESS is operationally integrated into the grid of the South Plains Electrical Cooperative (SPEC) and is controlled by SPEC with assistance from Younicos. During the course of the project, the BESS was thoroughly tested to determine its ability to optimize the reliability of the grid by compensating for the intermittency of wind, storing power during high wind- and providing power during low wind-conditions, reducing the daily peak load on the grid (and thus reducing demand charges), providing a ramping capability to stabilize the grid, and to potentially provide an ancillary service for fast response regulation service needed by the ISO to maintain the grid at 60 Hz. These and other potential uses of the BESS were exercised and evaluated as part of the project, and will continue to support its ongoing operation.

TTU researchers extensively modeled the battery system, the wind turbines and the related SPEC substation loads, and then simulated a full range of system behaviors involving a set of battery functions. The models and simulation results were then validated during field testing of the BESS during 2014. The basic approach for testing, describing such aspects as time of year, time of day, and duration, is outlined in the table below. Based on this, a detailed testing schedule was developed and implemented, involving single function tests and then finally combinations of functions, to determine the feasibility and economic viability of these functions. Some individual results of these functional tests are included after the table.

Table 1. Overall BESS Testing Approach

Number	Test Type	Time of Year	Time of Day	Duration
1	Ramping	All year	Night	Seconds
2	Demand Response	Summer (May to August)	Day	Minutes to hours depending on peak loads
3	Frequency Support	All year	Night	Seconds
4	Wind Speed Drop	All year	When drop occurs	Seconds to minutes
5	FRRS	All year	When needed	Until frequency recovers

One of the tests performed was frequency response. When the frequency went below 60 Hz, the battery automatically ramped up to supply the required power and met the demand. In the figure below, which demonstrates the responsiveness of the BESS, the black line is frequency (dead band is set to 59.70Hz), the orange line is system state of charge, and the blue line is kW output (with a maximum setting at 400 kW).

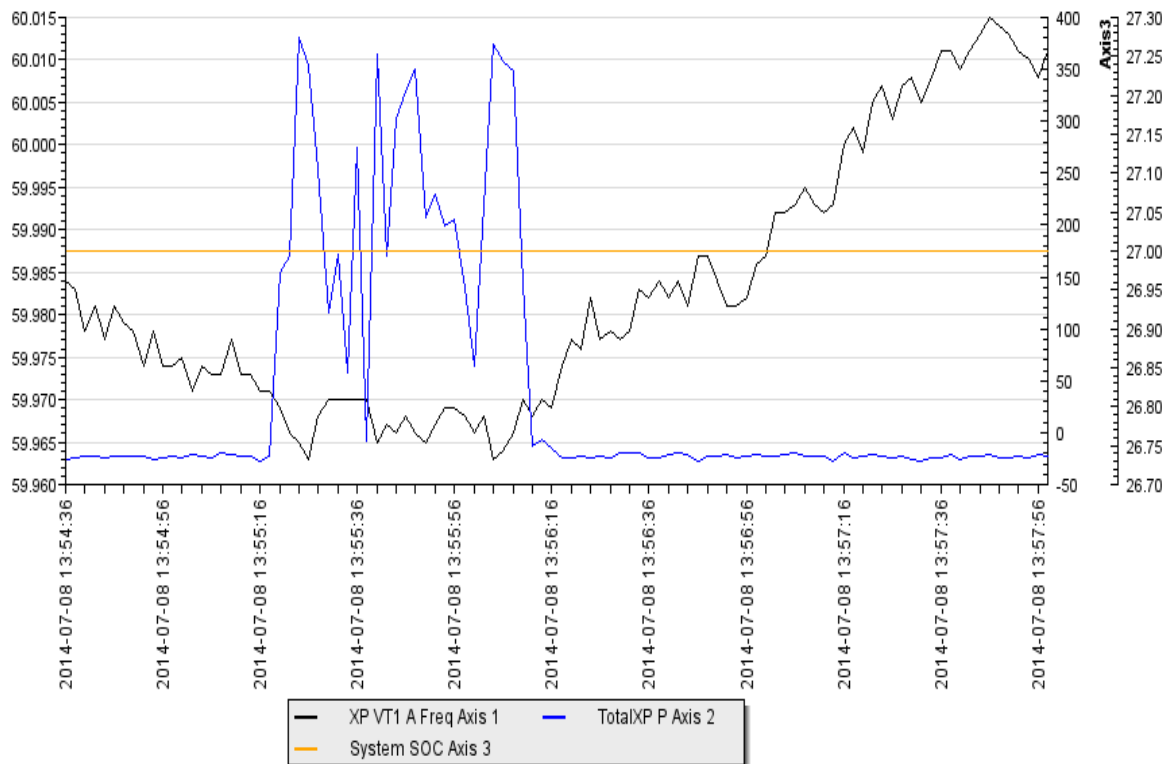


Figure 13. Under Frequency Event on July 08, 2014

Another core function of the BESS is to provide wind ramping/load leveling. The figures below show the battery following the wind profile according to the power generation and load demand. The green trace is the wind output for TTU-04 (Gamesa Wind Turbine) and the dark blue trace is the output of the BESS.



Figure 14. Load Leveling During August 2014

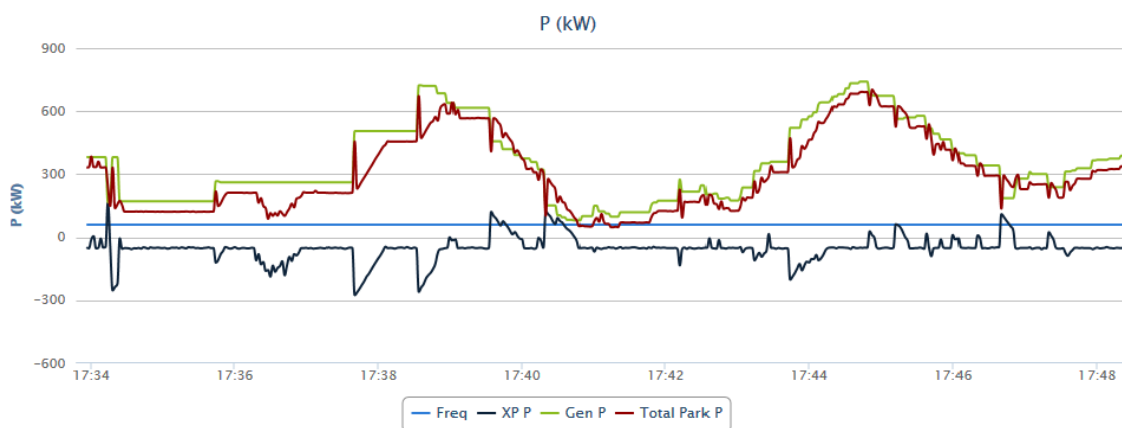


Figure 15. Load Leveling During October 2014

In addition to operating the battery in real time on SPEC's distribution system to learn the technical performance of the system, the team also evaluated the economics of an investment in a BESS. This involved two different activities: 1) development of a basic application to determine appropriate actions that increase profitability while considering battery life and 2) an economic analysis showing the return on investment using a DOE-supported evaluation tool known as the Energy Storage Computational Tool (ESCT).

For the first part, TTU developed an algorithm which weighed various factors. One of the most important factors is to determine the optimal time to charge and discharge the battery. The battery needs to be charged when demand and electricity cost are low and discharged when demand and electricity cost is high. It all depends on the purchase price of energy required to charge the battery and alternatively the sale price of energy to sell the excess power. The next important factor is to consider

the load/demand and supply of the battery power or the available renewable power in an economic way considering the past, present and future cost analysis. Based on the algorithm, TTU implemented a basic application that reads the input data from an Excel file every 15 minutes, and then through an interface presents a recommended battery action after checking seven different cases. An example case is shown below:

Case: If power generated from the wind turbine is less than the load demand, and the battery is in a charged state, then the condition “If present purchase price is greater than past energy to charge the battery” is checked. Then if the condition “If the future price is less than the current purchase price” is true, the recommended action is to “Discharge battery to load. Sell excess power. Charge battery in future.” as shown below. If this is not the case, then recommend “Discharge battery to load. Purchase power to supply load”.

The screenshot displays a software interface titled "Battery Recommended Action". It contains several input fields and a central action box. The inputs include:

- Time: 2013-12-01 00:00:00
- Battery Production (kW): 8.053333333
- Alstom Wind Production (kW): 7.4125
- Full Load (kW): 3481.75
- Past LMP Price: 55.0
- Current LMP Price: 60
- Future Projected Price: 45.0 (highlighted in green)
- Peak Shaving Limit (kW): 4500
- Battery Charge(0-100%): 100
- Current Grid Frequency (Hz): 60
- Peak Shaving Indicator: Below Peak Shaving Limit
- Frequency Indicator: Frequency Nominal
- Battery Charge Status(y/n): n

At the bottom left are three buttons: "QUIT" (red), "NEXT DATA" (grey), and "START" (green). In the center, a box labeled "Battery Action:" contains the following text:

Discharge Battery to Load.
Sell Excess Power.
Charge battery in future.

Figure 16. Algorithm Output – Discharge Battery to Load & Sell Excess Power

For the second part, the Team used the ESCT to perform a set of economic evaluations dependent on various factors such as the owner, type of market, location, and functional applications. To most fairly represent the economic potential in the Texas market, the Team performed four different sets of analysis, as outlined below. These sets represent various types of utilities, in both regulated and deregulated markets and located in different regions of the State, using a BESS to address diverse needs, from improved electric service reliability to ancillary services. It should be noted that the ESCT estimates of benefits primarily accrue to the owner of the BESS and do not easily represent multiple functions carried out simultaneously by multiple market participants.

The representative configurations of ownership, location and functionality include:

1. 1MW BESS on a distribution network owned by a utility in a deregulated market
2. 1MW BESS on a distribution network owned by a utility in a regulated market (without demand charge management)
3. 1MW BESS on a distribution network owned by a utility in a regulated market (with demand charge management)
4. 1MW BESS on a transmission network owned by a generator in a regulated market

Of these four options, only #3 proved to be economically viable, although the feasibility of other options will improve as the cost of a BESS is reduced. The following table shows the parameters given as inputs to the ESCT for option #3.

Table 2. Parameters for Asset Characterization Module

Location of battery	Distribution
Market	Regulated
Owner	Utility (End User)
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Expected year over demand growth of electric system	2%
Expected life time	20 years
Average inflation rate	2%
Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$2,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Primary application	Demand charge management
Selected secondary application(s)	Renewables energy time-shift and electric service power quality

The following table shows other data inputs for the ESCT.

Table 3. Review Benefit Calculation Inputs

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Average On-Peak Price of Electricity	\$/MWh	100%	100%	100%	\$ 100	\$ 100	\$ 100
CO2 Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
SOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	5.800	5.800	5.800
NOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.60	2.60	2.60
PM Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
Value of CO2	\$/ton	100%	100%	100%	\$ 20	\$ 20	\$ 20
Value of SOx	\$/ton	100%	100%	100%	\$ 520	\$ 520	\$ 520
Value of NOx	\$/ton	100%	100%	100%	\$ 3,000	\$ 3,000	\$ 3,000
Value of PM	\$/ton	100%	100%	100%	\$ 36,000	\$ 36,000	\$ 36,000
Average Monthly Load Reduction from ES at Peak	MW	100%	100%	100%	1.0	1.0	1.0
Average Monthly Load Reduction from ES at Partial-Peak	MW	100%	100%	100%	1.0	1.0	1.0
Average Monthly Load Reduction from ES at Off-Peak	MW	100%	100%	100%	1.0	1.0	1.0
Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 11,590	\$ 11,590	\$ 11,590
Partial-Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 2,650	\$ 2,650	\$ 2,650
Off-Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 6,890	\$ 6,890	\$ 6,890
Number of Months that Demand Charges are Avoided	#	100%	100%	100%	5	5	5
Typical Number of Power Quality Events per Year	#	100%	100%	100%	10	10	10
Value of Avoided Power Quality Events	\$/kw peak	100%	100%	100%	\$ 5.0	\$ 5.0	\$ 5.0
Customer Peak Load	kW	100%	100%	100%	1,000	1,000	1,000
Total Renewable Energy Discharged for Arbitrage	MWh	100%	100%	100%	782	782	782
Total Renewable Energy Discharged for Energy Time-Shift	MWh	100%	100%	100%	782	782	782

The next table shows the cumulative benefit and cost table, based on the reference case analysis, with the total cost of deployment and total net benefit shown on the bottom right of the table. As you'll note in these results, for the 20-year period, the ESCT estimated:

- \$2,309,600 in reduced electricity costs (mostly from reduced demand charges)
- \$594,900 in improved power quality savings (includes voltage and frequency support)
- \$546,600 in emissions savings

This total gross benefit of \$3,451,000 results in a return on investment of \$491,200 (after the cost of the BESS, annual O&M expenses, etc.) for this option (benefits/cost = 1.16). Note that without the the estimated environmental benefits, which are based on the future existence of a market for monetizing reduced emissions, the net return would be a negative \$55,400. Even under the favorable conditions of being able to serve all of the identified functions, a utility likely would not make such an investment until the regulatory rules allowed cost recovery of storage investments or at least the environmental benefits.

Table 4. Cumulative Benefit and Cost Table

	Cumulative Benefit and Cost Table			
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+ Primary and Secondary Benefits - Total Present Value over the Deployment Period	= Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -	\$ -	\$ -
	Capacity Market Revenue	\$ -	\$ -	\$ -
	Ancillary Services Revenue	\$ -	\$ -	\$ -
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -	\$ -	\$ -
	Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
	Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
T&D Capital Savings	Deferred Transmission Investments	\$ -	\$ -	\$ -
	Deferred Distribution Investments	\$ -	\$ -	\$ -
Energy Efficiency	Reduced Electricity Losses	\$ -	\$ -	\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -	\$ 2,309,600	\$ 2,309,600.00
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Interruptions	Reduced Outages (Consumer)	\$ -	\$ -	\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Quality	Improved Power Quality	\$ -	\$ 594,900	\$ 594,900.00
	Reduced CO2 Emissions	\$ -	\$ 127,500	\$ 127,500.00
Air Emissions	Reduced SOx Emissions	\$ -	\$ 13,900	\$ 13,900.00
	Reduced NOx Emissions	\$ -	\$ 36,200	\$ 36,200.00
	Reduced PM Emissions	\$ -	\$ 369,000	\$ 369,000.00
	Total Gross Benefit	\$ -	\$ 3,451,100	\$ 3,451,100.00
Costs				
	Capital Cost of Deployment (fixed charge rate)			\$ 2,406,800.00
	Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to loss- Decommissioning and Disposal Costs)			\$ 546,900.00
				\$ 6,300.00
	Total Annual Cost of Deployment			\$ 2,959,900.00
	Total Net Benefit			\$ 491,200.00

While only very special conditions such as those identified for option #3, including capturing the monetary value of emissions reductions, would justify an unsubsidized investment by a utility in a BESS, demonstrations like this one are critical to understanding individual utility, system-wide and societal benefits of storage. As battery technology improvements continue to drive costs down, and the regulatory rules get modified to more completely reflect the values of storage, battery systems will find their way into the mainstream of technologies for grids challenged by large concentrations of renewable generation.

Pricing Incentives for Reshaping Residential Load

Another key aspect of the set of demonstrations is variable electric pricing designed to measure the willingness of consumers to change behavior, by either reducing energy use, or in some cases shifting certain home activities to times when electricity is cheaper. The objective of this effort was to determine the degree to which electric customers would alter their pattern of electricity consumption in response to electric prices that vary with the time of day, as well as measure their response to different types of pricing programs. As an example, wind power is most prevalent during the late evening and night-time hours, and this project was one of the first in the country to leverage inexpensive wind energy and offer a special night-time wind-enhanced electricity price that was significantly cheaper than

day-time prices. Consumers that shifted their energy use from typical expensive hours (4 p.m. to 7 p.m.) to late evening hours (after 10 p.m.) were able to dramatically save on the expense of certain activities. Sixty-one homeowners voluntarily participated in the pricing experiments over the period of two years, and a like number of homes were unknowingly part of a control group for comparison.

Most of the participants in this study reside in a neighborhood known as the Mueller development which is situated about three miles from downtown Austin. The homes have energy efficient features, and the neighborhood has a high concentration of installed grid-tied solar photovoltaics (PV) and plug-in electric vehicles (PEVs). Basic demographic information and typical features of the homes are listed below:

Table 5. Demographic Data of Participants in Pricing and Control Groups

Demographic Information	Responses
Salary	43 survey respondents, or about 81%, claimed income above \$75,000. One respondent claimed income between \$300,000 and \$1 million, and two participants claimed income above \$1 million.
Ethnicity	51 respondents, or 82%, reported Caucasian ethnicity. Three respondents reported Hispanic ethnicity. Three reported Asian/Pacific Islander ethnicity. Five declined to respond.
Education	All respondents, except for one left blank, reported a college degree or higher. 42 respondents, or 71%, reported a postgraduate degree.
Home during weekdays	46% of the Pricing Group and 62% of the Control Group homes can typically be occupied on weekdays (daytime hours).
PEV ownership	34 plug-in electric vehicles (PEVs) are owned by the Pricing Group, consisting of 25 plug-in hybrids and 9 full electric vehicles compared to only 6 PEVs for the Control Group. Note - 34 PEVs in 62 homes is a much greater penetration than seen in the national market, providing a glimpse of potential future load shaping via shifting of vehicle charging schedules.
Cooking appliances	Predominantly natural gas. 27% have an electric oven.
Laundry appliances	1 in 3 has an electric dryer even though natural gas is available.
HVAC systems	90% have a gas furnace.
Water heaters	Natural gas
Average square footage	Approximately 2,000 square feet for both groups.
% of homes with programmable thermostats	100% of homes in both groups.

The homes within the Pricing Group and the Control Group were outfitted with home energy monitoring (HEM) systems. These devices attached to the home's individual electrical circuits within the breaker panel(s) and collected whole home and circuit (appliance) level energy use data in one-minute intervals.

This energy use data was used to perform comparative analysis between and within the Pricing and Control Groups to characterize and quantify the Pricing Group's response to two experimental pricing programs outlined below which did not coincide or overlap at any time:

- A seasonal time-of-use (TOU) pricing experiment featuring a discounted night-time price per kilowatt-hour to encourage use of available inland (west Texas) wind generation every day during the five windiest months of the year (March through May, and November through December)
- A critical peak pricing (CPP) experiment invoking an expensive price per kilowatt-hour during defined peak hours on no more than 15 designated "critical peak" days during the four summer months (June through September). Critical peak days were "called" a day ahead by the project team based on forecasted next day weather conditions under the premise that extremely hot weather might coincide with ERCOT peak loads.

A CPP price of \$0.64 per kilowatt hour (kWh) was only applied to the Pricing Group participants' energy consumption during the weekday (Monday – Friday) afternoon hours of 4 p.m. – 7 p.m. on "critical peak" days. To make the experimental pricing revenue neutral, there was an offsetting discount of approximately \$0.016 per kWh on energy consumption during all non-peak hours of CPP event days. Twelve and 15 CPP days were called in 2013 and 2014, respectively.

For the wind enhancement pricing trial, the night-time electric rate for the Pricing Group was lowered to \$0.0265 per kWh from 10 p.m. – 6 a.m. on all days during those months. To make the experimental pricing revenue neutral, there was an offsetting surcharge of approximately \$0.02 per kWh on the participants' energy consumption during all other hours of those five months.

Over the course of this 20-month experiment (March 2013 – October 2014), the Pricing Group and Control Group received and paid their usual Austin Energy electric bills, which reflected the normal Austin Energy electric rates. To aid in recruitment of participants, and to reduce turnover, this effort provided all participants with an incentive. This monetary incentive was used to entice the Pricing Group participants to either shift their energy consumption to the "wind enhancement" time period (10 p.m. – 6 a.m. during certain months) or to reduce energy consumption during CPP times (4 p.m. – 7 p.m. in the summer months). The project established a credit account with an initial balance of \$200 for each participant in the Pricing Group. The credit account balance at any point in time, whether above or below that initial \$200 amount, reflected the participant's degree of response to the pricing programs. If they reduced or shifted their use accordingly, then they actually accrued savings that were paid at the end of the experiment. Participants were able to track their progress and view their credit account through an online web portal display. For any participants who chose not to change their behavior, the pricing programs were designed to be revenue neutral meaning that participants who made no behavioral changes would not benefit nor be penalized. In fact, by the end of this effort, only 3% of the participants had deductions from their incentive account, and they were minor amounts.

Wind-enhanced pricing results

In the wind-enhanced pricing experiments, the results showed that homeowners are responsive to pricing systems that will shift their electric load significantly to assist the grid manager, and improve system efficiencies, by making the system load more compatible with wind generation. The figure below compares Pricing Group to Control Group daily consumption patterns during the 2014 March to May wind pricing months, plotted in hourly percentage of daily total usage. The seasonal average hourly Texas wind generation is the dotted line in the background.

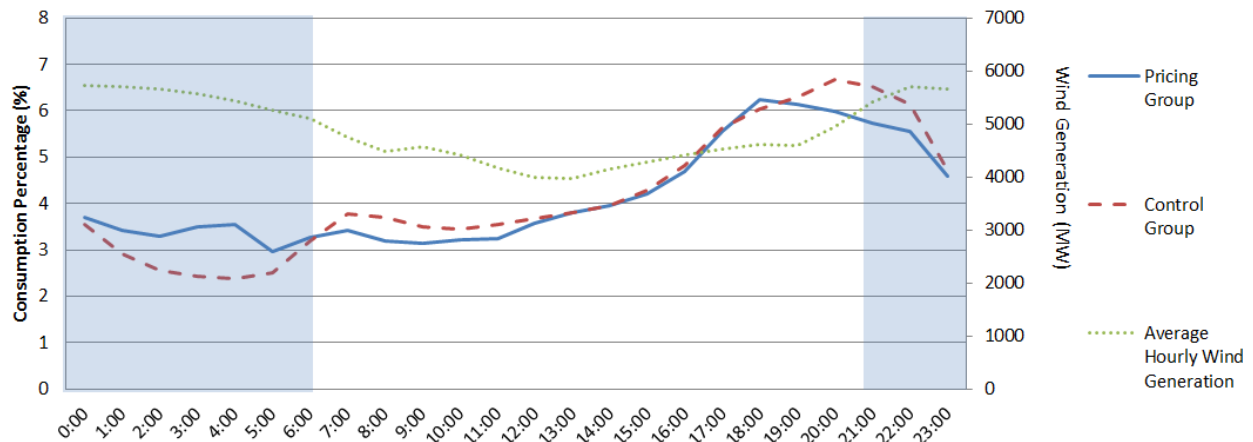


Figure 17. Pricing Group vs. Control Group 2014 March - May Wind Pricing Period

In contrast to the shifts evident in wind-pricing months (above), there is no clear shift in non-wind months (see graph below for June 2013). Energy consumption patterns for the Pricing and Control Groups are more similar in non-wind months.

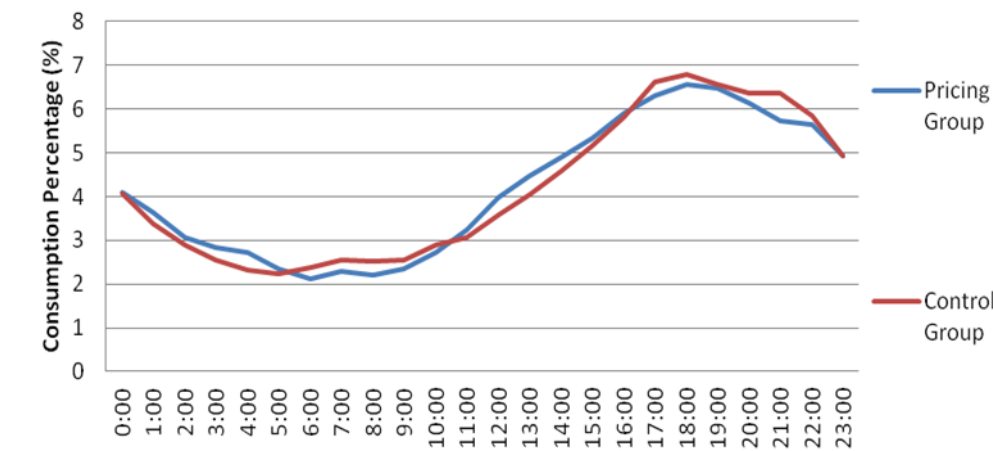


Figure 18. Pricing Group versus Control Group June 2014

Daytime/night time comparisons of overall electricity consumption for the Pricing and Control groups over all wind months during the experiment indicated a 4% difference, which is small but statistically significant considering what activities can be shifted to late evening or overnight hours (typically just PEV charging).

The average PEV charging consumption daily load shape is graphed below for owners of PEVs, for three different periods—before the trial started, during all wind months, and during the summer.

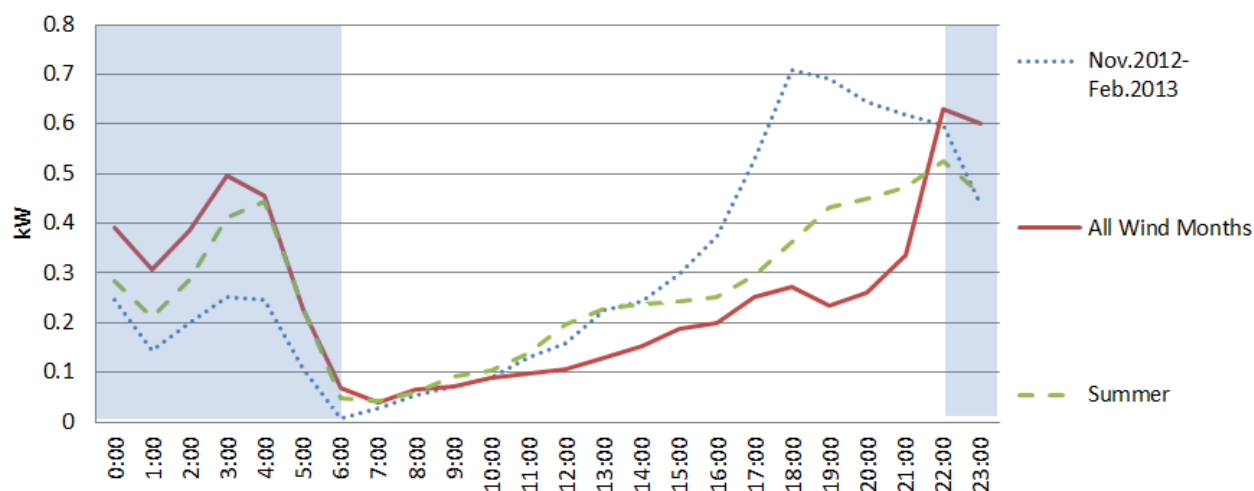


Figure 19. Pricing Group PEV Owners Average Charging Behavior

In the figure above, (blue dotted line versus solid red line) note the marked reduction in day time charging by PEV owners when the night-time wind pricing is in effect, taking advantage of very inexpensive overnight charging. Also note the partial rebound in day time charging during the summer months (green dashed line) when wind pricing is not in effect. This summer rebound may be partially caused by the longer daylight hours and increase in evening activities in the summer time, calling for increased seasonal afternoon charging. But some participants stated that they went back to charging when it was convenient (including daytime hours) since the wind pricing was not in effect.

CPP results

During the peak summer months, wind generation is very low so getting consumers to shift their load to the hours when the wind contribution is highest is not of much effect. So, a critical peak pricing program to simply encourage consumers to reduce or shift their load off the peak for a three-hour period (4 p.m. to 7 p.m.) is the most important for the efficiency of the ERCOT grid, and to make the grid manager's job easier when resources are limited. The CPP experiment showed that consumers will reduce their peak demand when they get a day-ahead notice and are provided an economic incentive to save money.

During the summer months of 2013, 12 CPP events were called. During 2014, that number was 15. The estimated kW savings achieved during those CPP events is shown in the tables below.

Table 6. 2013 Panel Data Analysis Results

Critical Peak Day	Daily High (F)	Estimated kW Savings
June 20th	98	0.5475
June 26th	101	0.7455
June 28th	104	0.7002
July 24th	97	0.4194
July 26th	99	0.3590
August 1st	103	0.2100
August 7th	103	0.7047
August 8th	104	0.6429
August 29th	99	0.2768
August 30th	102	0.2897
September 5th	96	0.1259
September 13th	97	0.3272
Average Savings per Event:		0.44 kW

Table 7. 2014 Panel Data Analysis Results

Critical Peak Day	Daily High (F)	Estimated kW Savings
June 24th	93	-0.2484
June 30th	96	0.2840
July 2nd	97	0.0068
July 14th	97	0.2064
July 24th	101	0.0052
July 28th	99	0.1574
August 7th	102	0.2292
August 15th	101	0.1420
August 19th	96	0.5093
August 21st	103	0.5163
August 26th	101	0.2264
September 3rd	100	0.1601
September 9th	97	0.2082
September 11th	99	-0.0185
September 15th	88	0.2618
Average Savings per Event:		0.17 kW

As evidenced in the tables above, CPP events in 2014 achieved less saving on average than in 2013. For comparison, the average reductions in demand were 0.44kW or 15.27% in 2013, but only 0.17 kW or 6.15% in 2014. From responses to the final survey of the Pricing Group, the greatest cause of lesser savings the second summer (2014 versus 2013) was that participants made lifestyle changes at the beginning or early in the trials that were persistent throughout the trial period, thus reducing the apparent 2014 savings yielded by panel data analysis and historical baseline method. Examples include moving their PEV charging, laundry and dishwasher use to outside the on-peak window. A lesser behavioral impact on 2014 savings was customer fatigue. In consideration of these performance factors, the 2013 results were considered more representative of the group's CPP response.

Further analysis for Pricing Group members with PEVs revealed that around 62% of the CPP kW reductions were attributed to HVAC and PEV related behavior change (HVAC @ 40%, PEVs @ 22%). For Pricing Group members without PEVs, around 57% of the CPP kW reduction was attributed to HVAC related behavior change. Apart from HVAC and PEV kW reductions, the other CPP kW reductions came from other uses with laundry and kitchen appliance consumption both found to be significant contributors.

Pricing Group members with EVs saved more energy during CPP events (0.46 kW versus 0.28 kW) than non-PEV owners, as depicted below for the 2013 events.

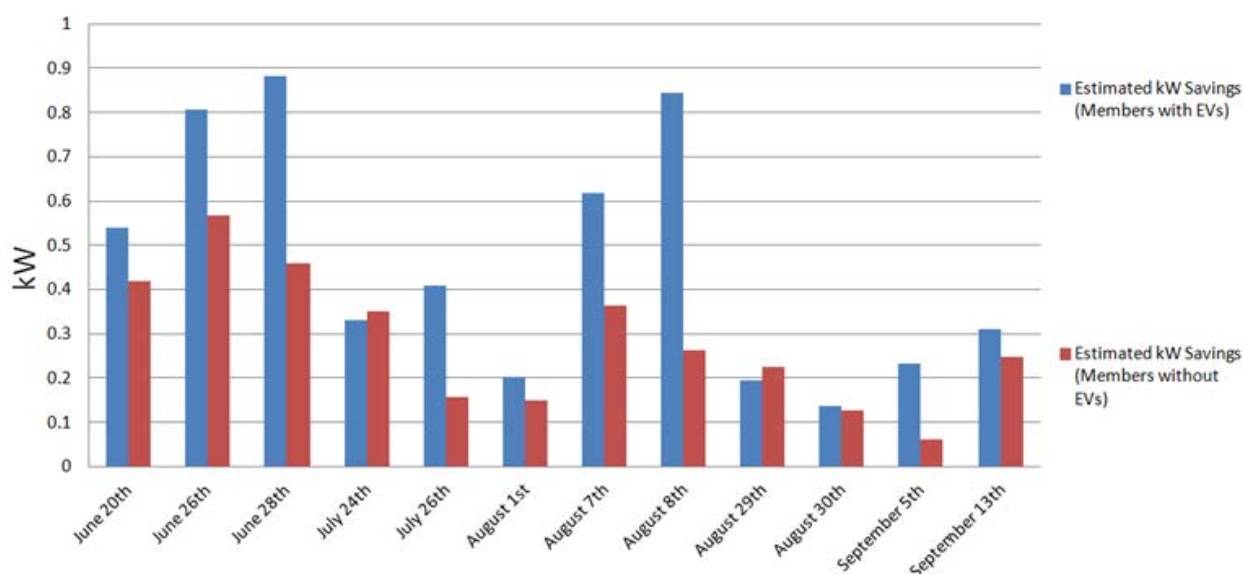


Figure 20. Comparing CPP Pricing Group Members with and without EVs

As expected, the Pricing Group participants who own PEVs exhibited more variability in responding to CPP events. If normal charging typically began when they arrived home, but they delayed that charging until after 7:00 p.m. during CPP events, then their savings were much higher than for those without PEVs. PEV charging is a high kW load that can often be delayed until later in response to pricing.

Although very successful, it should be noted that the Mueller community is not typical of communities throughout Texas, so one cannot analytically extrapolate these results to all of the ERCOT market. For perspective, however, if the 0.44 kW peak reductions per residence were fully realized statewide at Texas' 10,000,000 residential housing units, the system peak reduction would be 4,400 MW. If the wind-enhanced pricing trial were to be implemented statewide at every residential unit, successfully shifting 5% of load during shoulder months to night-time periods, the shifted load would equate to one half to two thirds of the current seasonally available overnight wind generation.

Solar Community-Grid Compatibility Challenges

Residential rooftop solar PV distributed generation is logically a good complement to wind resources on the grid since the solar contribution comes during the day while the bulk of the wind contribution (west Texas wind) is concentrated in the late evening and night-time hours. Concentrations of rooftop solar PV generation, however, bring their own challenges to the grid operator, and these are likely to be present during the shoulder months (March – June, and October – November) when the system demand is low, the sun is shining brightly, and west Texas wind is most abundant. The expected significant increase in solar adoption has raised a number of concerns for utilities, so the CCET project included a component to monitor and assess the dynamics of a solar community as it interacts with the grid.

One of the challenge areas resulting from increasing market penetration of residential rooftop solar PV distributed generation is variations in electrical performance parameters on residential distribution circuits and behind the meter, such as excursions in load, power factor and voltage, as well as harmonic distortion effects. As reported herein, these issues arise from multiple interacting and dynamic factors, including distributed solar generation cancelling utility kW load, variable solar PV generation, the increasing harmonic current drawn from modern home appliances, and others.

To address the key challenge areas of concern, the CCET effort included residential circuit monitoring at two levels: 1) on 35 kV underground residential distribution (URD) circuits in Houston, Texas, and 2) at the residential customer point of connection in Austin, Texas. The observations noted from these two efforts will help utilities anticipate the occurrence and assess the significance of potential power quality impacts in communities with a range of concentrations of solar PV and in making design and/or operational adjustments to mitigate such impacts.

For the first task, CenterPoint Energy led an effort to examine correlations between distributed solar PV generation and distributed circuit power quality impacts by measuring and comparing electrical parameters on a total of six 35 kV circuits in two neighborhoods in an area north of Houston, Texas, one with a moderate penetration of solar PV and the other with virtually no solar. The Harmony neighborhood (formerly Discovery at Spring Trails) has a high concentration of new homes (built in 2008 and after) with very efficient building features, energy efficient appliances, small (average 1.1 kW) rooftop or trestle-mounted solar PV systems, and HEM systems for monitoring electric and water use. The second neighborhood, Legends Ranch, includes similar type homes, but they are built without enhanced energy efficiency features and are not equipped with rooftop solar PV, energy efficient appliances or HEM systems, i.e. more typical of newer Texas neighborhoods. Specifically, this effort focused on analysis to compare and correlate solar PV generation to circuit parameters by measuring and calculating total harmonic current and voltage distortion (THDi and THDv, respectively), power factor (PF) and voltage at the distribution level as 5 minute averages in the two neighborhoods.

For the second task, a total of seven homes in the Mueller neighborhood in Austin, Texas, were monitored by Pecan Street for power quality parameters, solar PV production and net load served by the utility. As mentioned before, the Mueller area is a relatively new, “green built” neighborhood with

efficient home construction, energy efficient appliances, a high concentration of installed grid-tied solar PV and plug-in electric vehicles. This neighborhood has approximately 30% solar market penetration and the average installed system nameplate DC capacity is 5.24 kW for each home having solar PV. In general, this investigation provided an opportunity to develop a more complete understanding of the interaction of solar PV with the distribution infrastructure. The root cause of any observed issues should be traceable to events occurring within the residential load, residential solar PV operation, or the utility distribution service. Information gained on the secondary side of the service transformer can be compared with measurements at the distribution circuit level (even a different circuit) if the load type and distribution between the two circuits are similar. Previous studies have shown residential power factor to be .9 to .95 lagging, and typically of low distortion. The addition of solar PV and the proliferation of switch mode power supplies are changing these characteristics. The benefit to the utilities is an updated and more accurate characterization of the residential electrical load at the point of connection.

The homes monitored for this study were selected because they share two 50 kilovolt-amperes (kVA) distribution transformers, and because they are participating in Pecan Street research trials, whereby the residents have agreed to allow their energy usage to be monitored. The two transformers for the selected homes were monitored allowing the data to be compared with measurements on individual homes. The transformers were loaded as follows:

- Transformer 7007 has 10 connected houses, 3 with solar (30%); 18,800 square feet total
- Transformer 7006 has 8 Houses, 5 with solar (63%); 14,200 square feet of total

The solar monitoring data and analysis results identified a particular challenge of harmonic distortions from homes with solar during certain time periods, conditions of concern to the distribution utility. These harmonics can cause problems like degradation of conductors and insulating material in consumer appliance motors and utility transformers. Therefore it is important to gauge the total effect of these harmonics.

The observations of harmonic distortions in the DAT study were most prevalent in the shoulder months (March to May, and October to November in Texas) when home load is small (little or no AC load) on a sunny day when the solar unit is producing maximum power. Note that the solar unit from 10 a.m. until 4 p.m. is providing most of the power for the home and the grid-provided power is very low. The general conclusion of the study is that for homes with solar capable of providing all of the home's power use during a sunny day in the shoulder months, the harmonic distortions will be high, as shown in the two figures below which compare Harmony (solar) with Legends Ranch (no solar). The first figure represents a sunny day in April when solar production is able to cancel all but 20 kW of the entire utility load (60 homes). In contrast, the second figure is absent the solar PV influence so the THDi on the circuit varies throughout the day, and is lower in the daytime even though the kW load on the circuit is high.

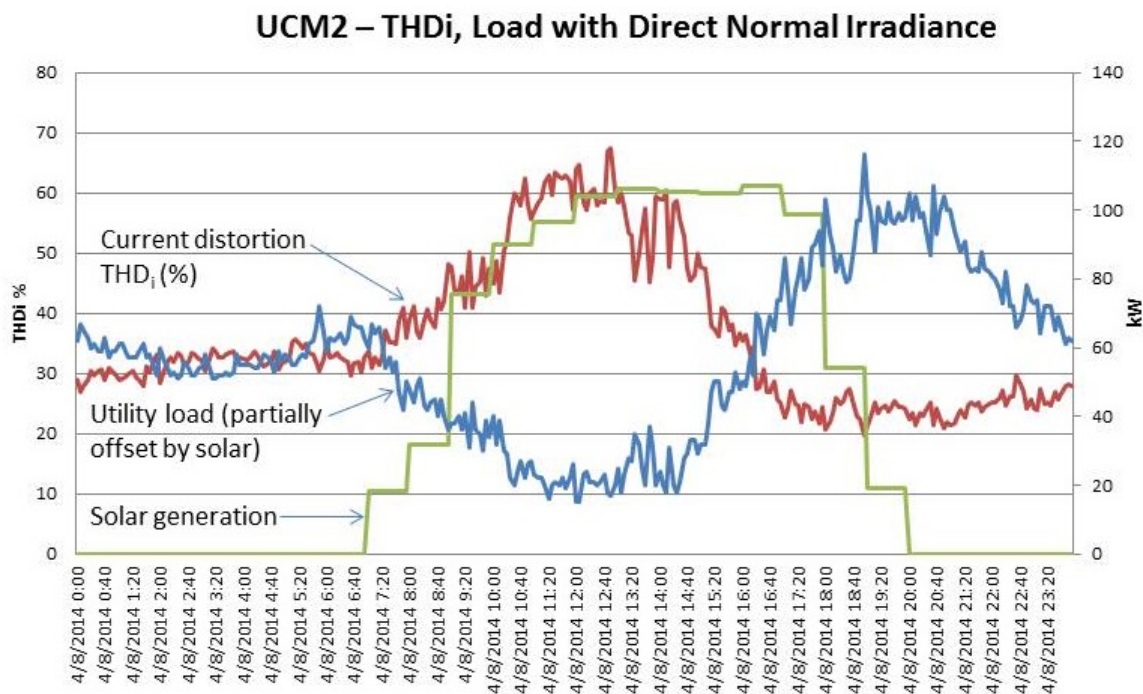


Figure 21. Movement of Harmony THDi, Solar Irradiance and Utility Load on a Sunny Day in April

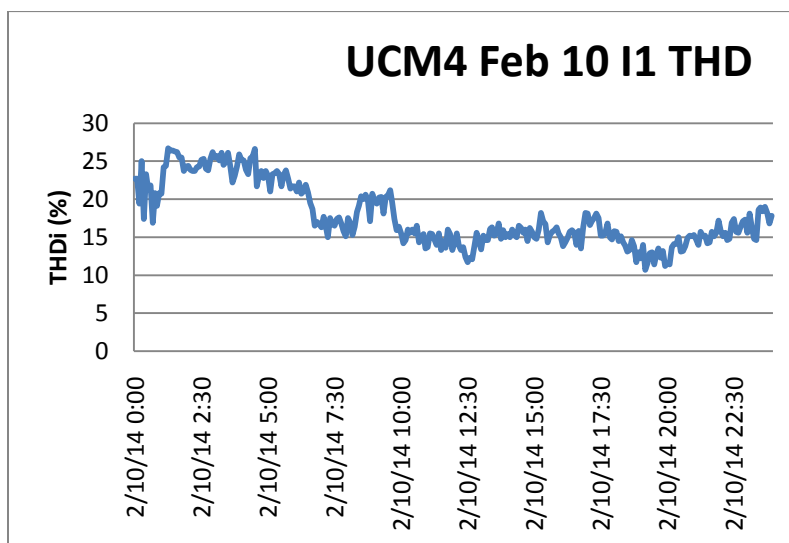


Figure 22. THDi Movement in Legends Ranch

To determine whether the THDi and PF behavior observed on the 35 kV URD circuits in Houston was unique to that neighborhood, similar monitoring and analysis was performed on the homes in the Mueller neighborhood in Austin. The same type of THDi and PF behavior as reported in Houston was observed in the Austin residential monitoring. For the Austin houses with larger solar arrays, longer periods of fundamental load cancelation were observed, though this will vary by day, weather, array size

and home electricity usage. This behavior will likely be seen at some point during the year by any house where solar PV production can offset an appreciable portion of the total home kW load.

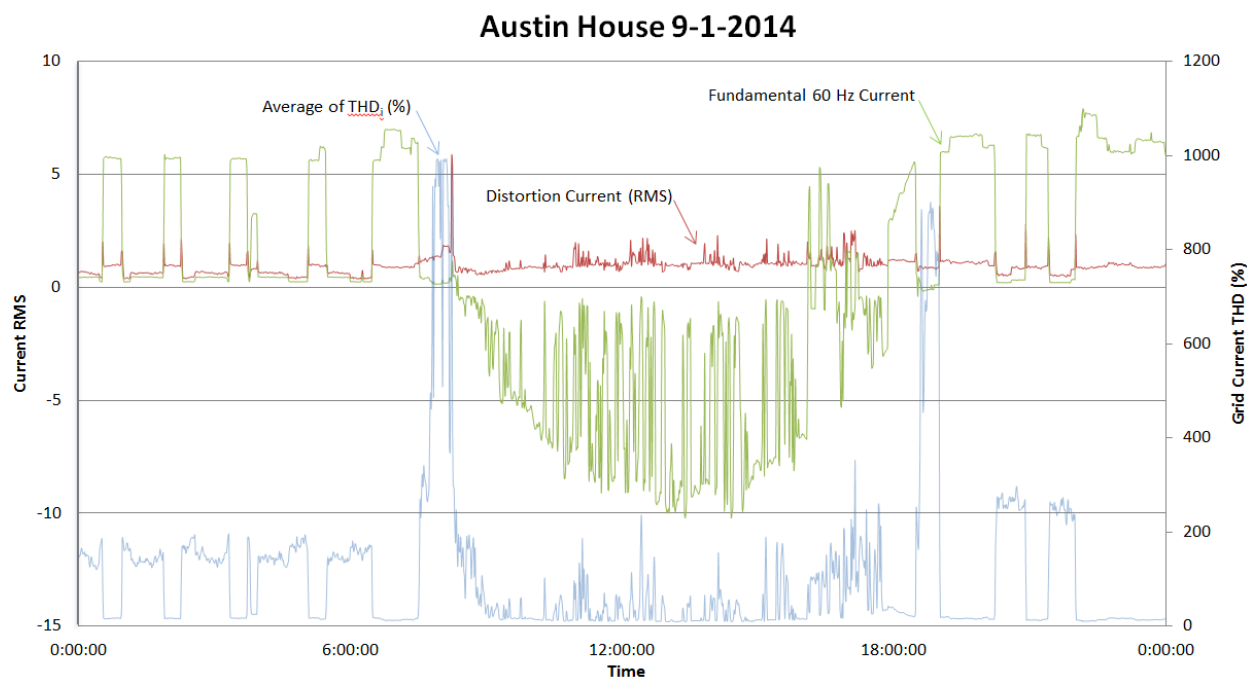


Figure 23. THDi Behavior for Home with Larger Solar PV System

In the figure above, the distortion current stays approximately the same during the entire day with variations that can be attributed to device usage, not solar production. During the daytime, solar capacity exceeds home load much of the time, driving the load on the utility (green line) negative, with oscillations due to cloud cover and home appliance operation (generally HVAC). During periods when the utility load (green line) is canceled and approaches zero for minutes to an hour at a time, the THDi spikes are very high, up to 1000%. This is due to the very low amounts of fundamental current supplied by the grid, not the solar production adding distortion to the grid.

Other findings of interest or conclusions include the following:

1. As the installed price of solar PV continues to decline, there will be continued increases in market penetration and average PV system capacity. Many homes already have several kW of installed solar PV and these homes produce their own real power (kW), which at times cancels the real power supplied from the utility. To the extent this condition exists in a given neighborhood, the customer at times would be drawing only distortion current and reactive power from the utility, leading to high harmonic current distortion (measured as THDi) and potentially a sagging power factor on distribution circuits.
2. It was shown that measured THDi can be significantly higher on distribution circuits that feed high concentrations of homes with solar PV as compared to those serving homes with little or

no PV. It is not clear that increased THDi is an immediate threat to reliability, but a review is suggested of the underlying assumptions of IEEE-519 regarding the damping effects of resistive loading. When the resistive loading is canceled by solar production, the impact of harmonics on grid stability may become an issue of concern that is not addressed by IEEE-519.

3. The power factor (PF) variability on distribution circuits with high solar PV penetration will likely require enhanced monitoring and upsized (and fast acting) PF correction to meet operating standards. One means of mitigation could be volt-amp reactive (VAR) injection at the solar PV inverter, and this could require compensation to the PV owner (customer) for any added equipment cost and for kW production forfeited in producing the injected VAR.
4. The localized voltage boost associated with residential PV production is additive at the service transformer and will propagate to distribution voltage. The resulting boost on the distribution circuit should not be problematic if the magnitude and rate of voltage drift can be handled by the distribution transformer tap changers.
5. All houses on a common service transformer will experience boosted voltage from PV production on any one or more of the connected houses. The boosted voltage will vary at rapid ramp rates due to fast moving cloud cover affecting PV production. The magnitude and potential impacts of limiting case voltage changes at service voltage and the need for and means of local monitoring and mitigation are unknown and should be defined.
6. PV inverters conforming to industry standards do not add appreciably to voltage or current distortion.

An interesting suggestion from this analysis was the potential benefit of using distributed energy storage systems placed either behind the meter or in the utility infrastructure. Such storage could allow for solar energy to be collected and stored for use throughout the day, minimizing power quality impacts by smoothing out rapid PV production ramps and reducing the time that the residential load is entirely canceled by distributed generation.

Direct Load Control Enhancement via Dual Communication Paths

There are times when the energy demand may come close to exceeding available generation resources. At times like these, the ability to call upon other resources to reduce the demand is imperative. One such action is termed demand response (DR) whereby commercial and residential resources are tapped to reduce their energy use. A need for demand reduction for grid reliability may occur during the summer peak (e.g., 4 p.m. to 7 p.m.), but it is also likely during the shoulder months of March to May when wind resources provide a large percentage of generation. In fact, the most recent load control event which included rolling interruptions in ERCOT occurred not during a summer peak but during a morning peak in February 2011. That event occurred when extremely cold weather raised demand and 5,000 MW of gas and coal units went off line. The wind was blowing and producing about 3,500 MW of wind power, but there were no other immediate resources available. Under these circumstances, demand reduction from large industrial consumers saved the day, but it could have easily come from thousands of homes as well.

In the residential sector, some homes are equipped with technologies, such as smart programmable thermostats, that can be remotely controlled (direct load control) by utilities or retail electric providers (REPS) to reduce the amount of energy use when needed. The consumers are incentivized, either with free equipment to help them better manage their energy use and costs, or through some type of small monthly payment. In many cases, the remote access is via the Internet using the homeowner's broadband network to control the equipment. Although this approach is technically feasible and fairly easy to implement, remotely contacting and controlling the equipment in the participating homes during a demand response event has been problematic. For a number of reasons, the utility or REP has not been able to access a high percentage of these devices during critical times, lessening the effectiveness of these programs especially considering the costs involved. For example, one of the Texas REPs, TXU Energy (TXUE), has provided home area networks (HANs) comprised of programmable thermostats and gateways to thousands of customers with the intent of leveraging these devices during demand response events. Based on a historical online percentage, for every 10,000 thermostats participating in the TXU program, only about 55% were able to be remotely accessed and cycled. Presuming each air conditioning system used an average of 1.5 kW, then the remaining 45% of offline devices equated to about 6,750 kW of load that could not be reduced during the critical time. To address this challenge, the CCET project undertook an effort to demonstrate that these results could be improved by leveraging other technologies to deliver the DR messages via a second communications path. This approach included the transmission of messages to the thermostats by using the transmission and distribution utility (TDU) networks and the residential smart meters, simultaneous pairing of the thermostats with both the Internet and the residential smart meter so they could receive messages using both paths, and use of the Smart Meter Texas Portal (SMTP) as a venue for distributing the messages to the utility networks. The figure below illustrates the high-level proposed architecture for this project.

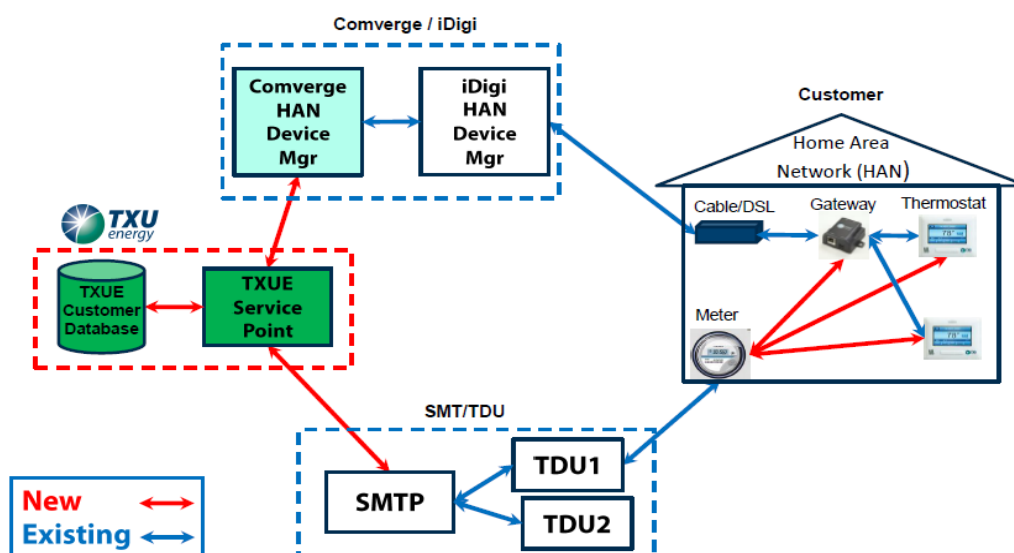


Figure 24. Proposed Architecture for Improving Demand Response Results

TXUE had thousands of homes already equipped with programmable thermostats, so the first solution was to remotely upgrade the software controls in these thermostats so they could also accept a signal through the home's smart meter. The technical laboratory and field tests of the dual-path system established the viability of employing both communication paths to improve reliability of communications while preserving the important ability to maintain a rich customer experience via information exchange supported by the Internet-based system. However, the prospect of completing the upgrade remotely over-the-air failed on the legacy equipment. Also, it was determined that many of them were not compliant with existing standards for communicating with the smart meter. The Team then evaluated two other approaches, but in each case there were only two viable options: 1) roll a truck and crew to replace the existing legacy equipment, or 2) send the equipment to the residence and have the homeowner install and configure the new equipment. The second option seemed viable, so tests were run using a friendly homeowner. Although the home testing proved that everything could be accomplished, and the DR messages could be sent and received via both paths, the testing also demonstrated that the actual equipment configuration process was fairly burdensome and problematic for a typical homeowner. Therefore, it was decided this option would not work. The first option, dispatching a crew, was deemed to be too expensive, so the project was halted. Therefore, while the dual path technology option is practical with new devices that are compliant with the standards, the retrofit of legacy equipment would require an expensive truck roll that could not be justified or financially supported by the project.

Aside from evaluating and demonstrating the dual path approach for improving DR, the project also explored the economic viability of DR in the ERCOT market. Since 2008, TXUE has initiated DR events across 13 days, including five pre-season tests. TXU Energy currently deploys a 15-minute, 100% cycling strategy, and the customer topology allows for TDU segmentation as well as sustained cycling within a region by rotating customer groups across 15-minute intervals. This approach allows TXUE to respond quickly to initial grid conditions with a predictable load reduction, and also allows for flexibility when extended cycling is required.

In order to estimate the average customer load reduction achieved during a DR event, 15-minute customer meter data was pulled for all days where DR was executed in 2011 and 2013. The figures below reflect a portion of the actual DR results based on a sample size of ~6,000 total customers. The first figure shows the effects of a single DR event triggered at 4:55 p.m. (not fully aligned with the 15-minute interval data). The subsequent set of figures shows the effects of one or more events triggered over several days in August 2011.

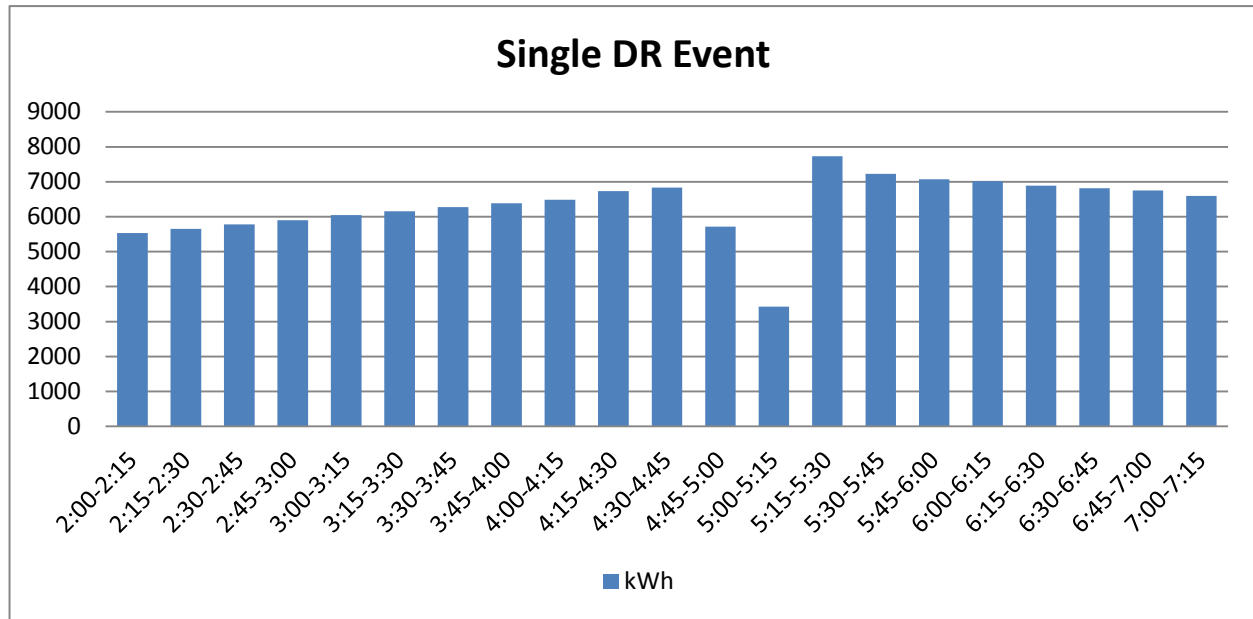


Figure 25. Single DR Event

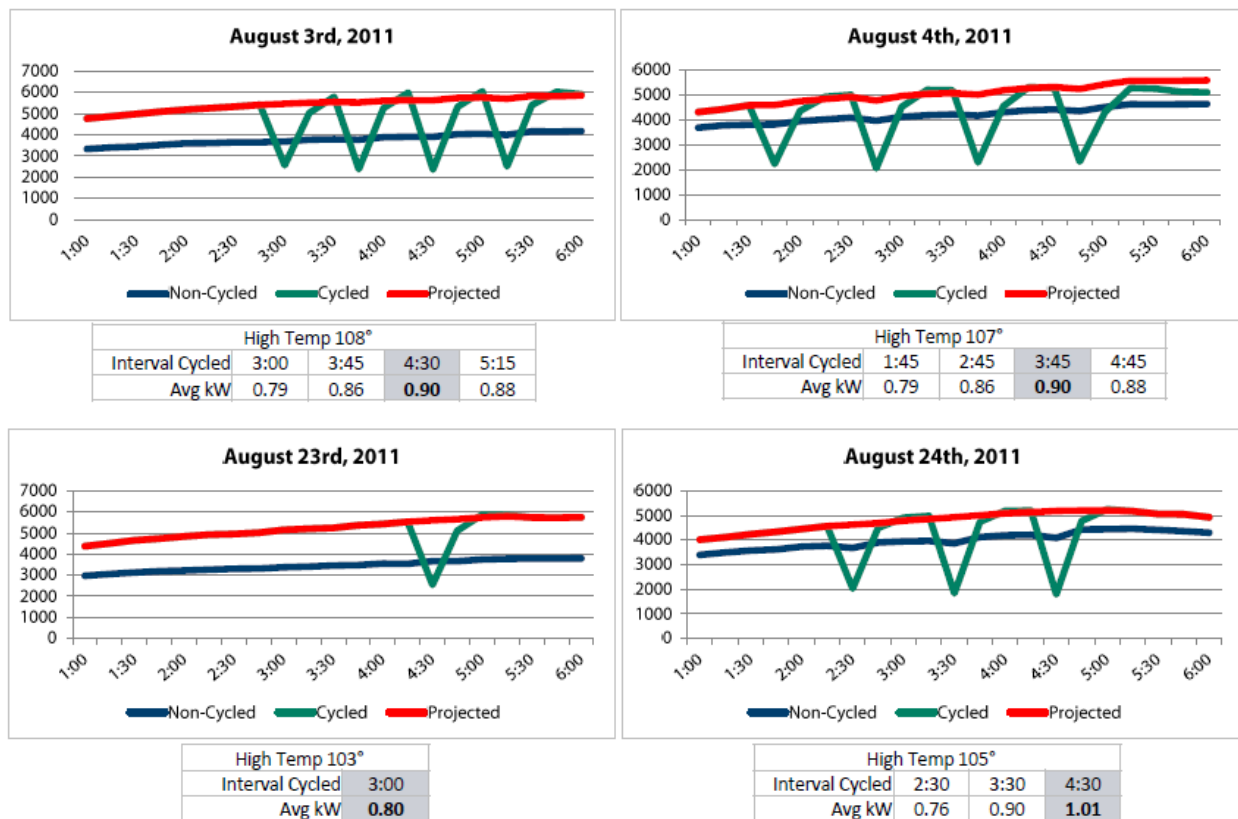


Figure 26. Multiple DR Events

Analysis of the 13 cycling days reveals the average reduction per customer (difference between cycled and smoothed) ranges from 0.50 kWh to 1.0 kWh. The variation can be directly correlated to outdoor temperatures. When cycling was performed on days where temperatures were in the upper 90s (not shown above), the average reductions ranged from 0.55 kWh to 0.75 kWh per customer. As temperatures climbed above 100 degrees the average reductions ranged from 0.75 kWh to 1.0 kWh per customer.

These averages can be used to determine the incremental value of controlling additional thermostats. From a pure cost perspective, the following chart represents the potential annual value of managing peak load at the maximum ERCOT bid cap levels. The table shows results for a suggested cycling of 1,000 homes for 15 minutes (maximum value per DR event) and then multiplies that by 10 DR events per year to derive the maximum annual value per 1,000 customers per year.

Table 8. Annual Value Proposition

Bid Cap (\$/MWh)	\$5,000	\$7,000	\$9,000	\$5,000	\$7,000	\$9,000
Less standard cost (\$/MWh)	\$90	\$90	\$90	\$90	\$90	\$90
Opportunity value (\$/MWh)	\$4,910	\$6,910	\$8,910	\$4,910	\$6,910	\$8,910
Average load shed per home for 15-minute event (kWh)	0.55	0.55	0.55	1.0	1.0	1.0
Customers cycled	1000	1000	1000	1000	1000	1000
Maximum value per DR event	\$2,701	\$3,801	\$4,901	\$4,910	\$6,910	\$8,910
# of DR events per year	10	10	10	10	10	10
Maximum annual value per 1,000 customers	\$27,005	\$38,005	\$49,005	\$49,100	\$69,100	\$89,100

While these results can be extrapolated for a much larger DR pool of homes, the respective cost of equipment, truck rolls, and support must also be extrapolated. For example, in the table below, if a REP determined that they could outfit a customer with a gateway and PCT for \$300 (including equipment, truck roll cost, and ongoing support), it could take 6-10 years (depending on load shed amount) to recover that cost at the \$5,000 bid cap, and 4-6 years at the \$9,000 cap. Given that Texas is a competitive retail market, and customers can swap REPs at will, equipping them all with these devices with the primary intent of supporting DR would be a huge investment with equally great risk.

In summary, the project was unable to demonstrate the ability to improve DR results by sending the DR signals via two paths to a large number of homes. However, the project did prove that it can be accomplished so one could estimate that the REP could potentially reach another 35% of their residents, meaning only 10% of the population would be unreachable. Unfortunately, as shown above in the economic analysis, the amount of time (in years) that it takes to recover the initial up-front investment

makes it difficult to justify the costs simply to provide a DR capability. Certainly the dual path approach should be considered in defining a workable architecture that can support DR, and especially if the equipment investment is going to be made by the REP anyway. Such a solution could be an important contribution in future DR program approaches where the dual path capability adds only a small incremental cost to the set of items needed to equip and engage the customer.

Fast Response Regulation Service using Fleet Electric Trucks

Frequency regulation is a grid ancillary service (AS) that allows electricity generators, including those with large battery energy storage systems, to bid services into the energy market to assist in maintaining the electric grid frequency at 60 Hz. Providers of this type of service guarantee to be able to adjust generation during a specified time period by either ramping up or ramping down production according to instructions in order to maintain the balance of generation/load on the grid, and thus the frequency. Traditionally, in the ERCOT market, this was accomplished through the dissemination of a four-second signal, with each frequency regulation generation resource then responding within their pre-established ramp rate until they met the full capacity specified in the signal. Participants in the frequency regulation market get paid whether they are called upon to adjust power or not, thus they are paid to be “on standby”. As shown in the figure below, ERCOT studies (data is from 2008) have determined that the need for fast response resources increase as the concentration of wind grows on the grid. This CCET project was undertaken to demonstrate the capability of addressing frequency regulation through the use of aggregated electric vehicle (EV) fleets.

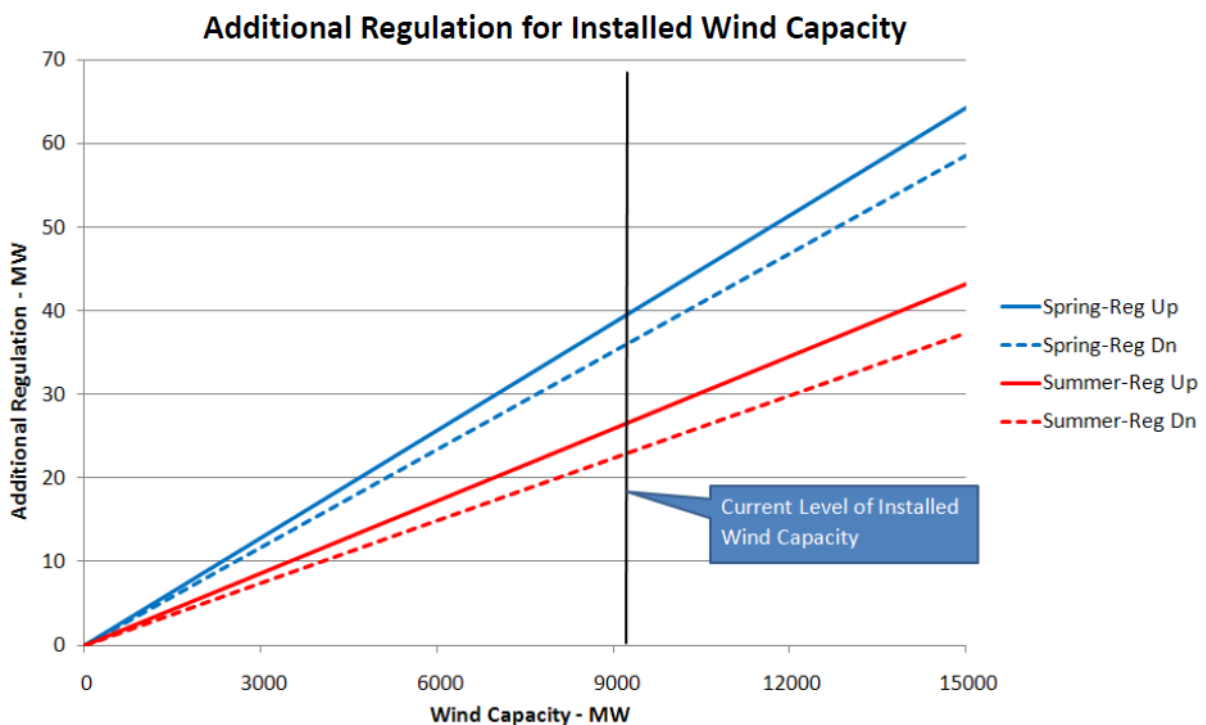


Figure 27. Results of a GE Study on Regulation Service

While the CCET DAT project was ongoing, ERCOT carried out a pilot project to find qualified resources willing to participate in an AS market called Fast Response Regulation Service (FRRS). Entities within ERCOT who could meet the technical requirements (primarily to respond with the promised kW of capacity within one second) were invited to participate. The purpose was to demonstrate that FRRS providers could improve ERCOT's ability to maintain grid frequency at 60 Hz when needed. The response requirements for this FRRS pilot basically limited participants to those with energy storage systems (batteries). Two large battery systems participated in the ERCOT pilot, and CCET seized on the opportunity to demonstrate that FRRS could also be provided by aggregated electric vehicles (EVs). CCET recognized that a great resource for batteries that could serve this purpose without significant investment (such as a large-scale utility scale battery costing millions of dollars) is through the use of aggregated electric vehicles (EVs). In particular, supporting regulation-up services (the reducing of load on the system) could be as simple as "flipping a switch", i.e. turning the EV charger off. Regulation-down (the addition of load on the system) could be provided almost as quickly although the delivery of load to the grid from the batteries would require a more extensive management and control system. CCET participated by qualifying, placing bids, and delivering the needed FRRS in the AS market with a resource derived from a fleet of electric trucks.

Electric delivery trucks are new technologies that are finding their way into special markets where the travel requirements, convenience and work schedules make the economics work. But as is usually the case, the economics are not currently favorable without government subsidies. However, the availability of being able to provide an additional service that generates revenue while the vehicles are stationary and essentially not being used would help the return on investment.

The CCET project leveraged the resources of Frito-Lay in Fort Worth to provide the electric vehicles while other team members developed the software and hardware capability to aggregate a set of 11 Frito-Lay trucks (shown below) so they acted as one resource capable of providing at least 100 kW of FRRS. The Frito-Lay trucks typically begin deliveries at about 3 a.m., return to home base about noon, and remain at that site for battery charging and loading of the next day's delivery until the next morning. This schedule is an ideal match for Frito-Lay's participation in the AS market since they were available to deliver the service when it was needed, and doing so would not detract from the primary duty of their trucks.



Figure 28. Frito-Lay Electric Truck and Charging Stations

This effort entailed the development and demonstration of an EV aggregation control system to support regulation-up services in response to an ERCOT signal or a detected deviation of grid frequency of greater than .09 Hz. The architecture for this effort is shown below.

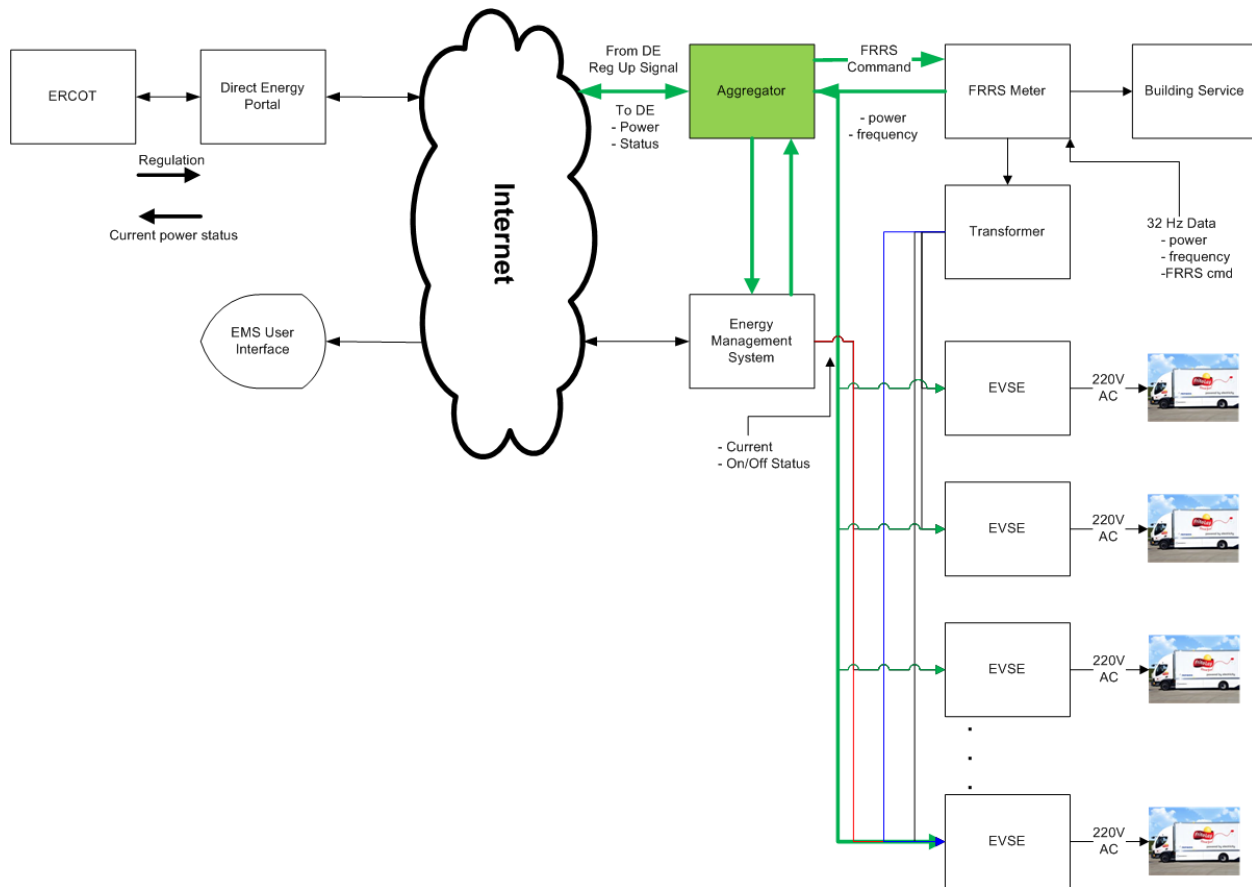


Figure 29. Architecture for Delivering FRRS from the Frito-Lay Facility

The following activities were conducted to create and ensure a successful deployment of technology:

- Identification and procurement of a utility revenue-grade meter to monitor power and grid frequency at a minimum of 32 Hz. After an extensive trade study of suitable meters to meet the requirements of the electrical infrastructure at the site, it was decided to build a custom power meter. The electrical infrastructure would have required either a complete replacement of the electrical panel box to split off building power from charger power or the acquisition of 12 revenue-grade power meters. Designing and building a custom meter that could monitor the 12 EV circuits after the panel box became the more cost-effective solution using National Instruments components. Total hardware costs were under \$20,000 versus over \$100,000 that would have been spent on individual power meters or redesigning the electrical panel box. The final solution, shown below, resulted in a compact, robust, fast, and inexpensive power

meter/aggregation system package that is easy to install and remove without requiring extensive site improvements to the electrical infrastructure.



- Development of a software aggregation control system to manage an EV fleet in accordance with FRRS parameters.
- Establishment of a secure physical network interface to the ERCOT system in accordance with FRRS pilot requirements. In accordance with ERCOT rules, this interface was done through a qualified scheduling entity (QSE). A DNP3 interface to the QSE was implemented over a secure VPN tunnel. The QSE telemetered the information provided by the aggregation system to ERCOT on a 2-second interval. ERCOT verified they were receiving the telemetered data from the aggregation system.
- Development of the data collection, correlation, and delivery to ERCOT of the required meter data files within one day of each frequency regulation bid period. The eventual solution to this was somewhat manual due to the rules for a QSE. After each bid hour, data files were automatically downloaded through a remote agent to SwRI. The data files were converted to a Microsoft Excel format, and then emailed to the QSE. After review, the QSE then sent the files to ERCOT. This solution worked well for the pilot and could be further automated in a future deployment.
- Development and deployment of hardware for controlling the EV chargers within the FRRS established time frames for power responses.
- Installation and deployment of secure networking hardware for the interface between the QSE and ERCOT.

Leveraging these capabilities developed and provided by CCET, the EV fleet at Frito-Lay successfully bid into the ERCOT FRRS market and received compensation for providing energy capacity to the grid. This became the first time EVs have successfully participated in an ancillary service in ERCOT and the first time a fleet owner has received payment for providing these types of services using an EV fleet. The

project results demonstrated that for the first time in the U.S. the battery capacity from a fleet of vehicles could be aggregated as an effective resource for grid management, and could do so without detracting from the primary vehicle purpose. Such resources in the future have appealing qualities for a number of reasons, especially in that they can provide such services to the grid operator from disparate locations, and in varying quantities; truly a distributed grid resource.

Although the project was highly successful, the real issue is the economic value proposition since the goal was to define an additional revenue stream that would help justify the costs of electric trucks. In the ERCOT market, given the current market rules, and with energy being relatively inexpensive in Texas, there is little to no economic benefit that can be realized by small fleet owners participating in the FRRS market. As a part of this project, the Team looked at ancillary service prices paid for frequency regulation services in ERCOT for 2013 and up through March 2014. Average frequency regulation prices paid per 100 kWh, by hour, for 2013 are shown below.

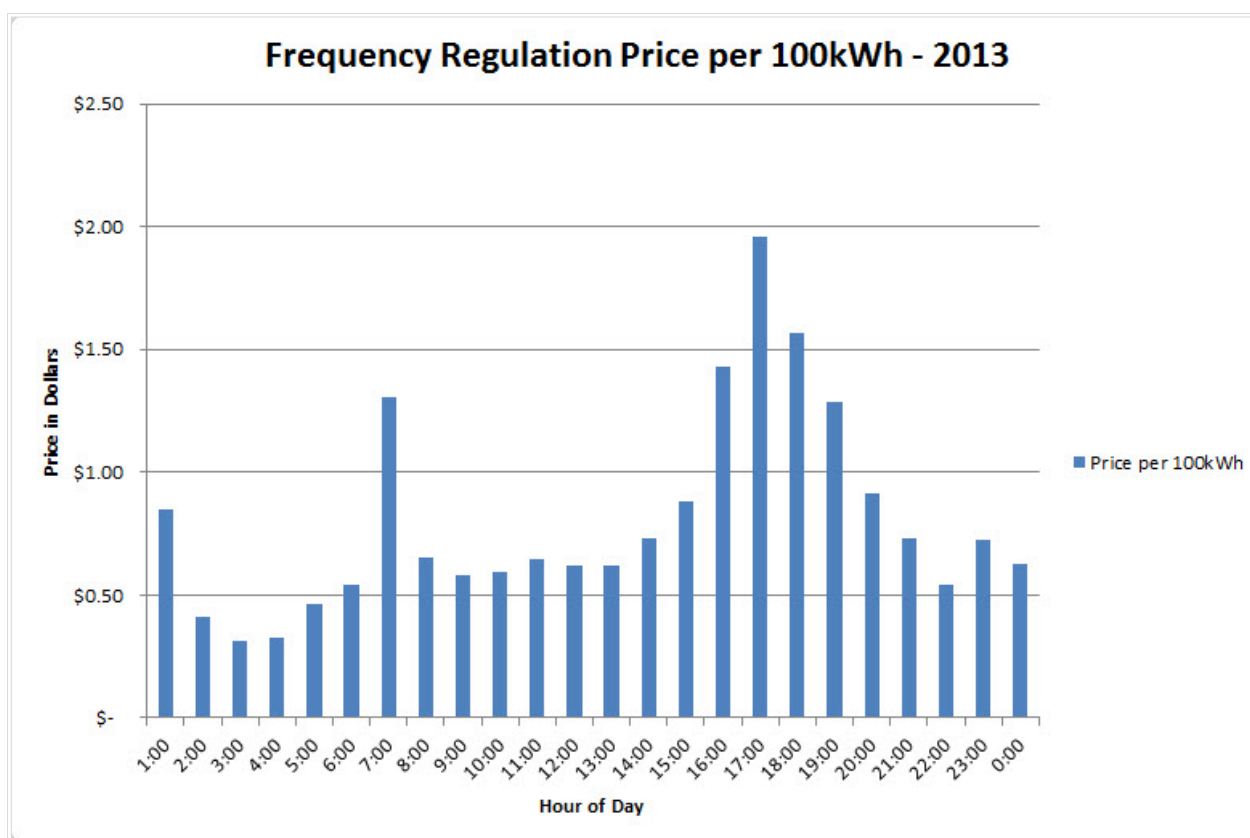


Figure 30. Average Frequency Regulation Price Per Hour Per 100 kWh in ERCOT for 2013

In 2013, there was a specific peak in frequency regulation pricing in the 6:00 a.m. - 7:00 a.m. time period, and a broader peak from 3:00 p.m. - 8:00 p.m. For the fleet in this project, the broader peak aligns well with the typical idle period of the vehicles, which do their deliveries between midnight and 10:00 a.m. and are loaded during the afternoon hours. Starting at around 5:00 p.m. they are all sitting on the chargers loaded and waiting for their next delivery run. The average price per 100 kWh,

however, is relatively low, around \$.80 - \$1.95 per 100 kWh. Looking at a potential of 4-hours per day that the project fleet could realistically and reliably participate in the FRRS market (from 5:00 p.m. – 9:00 p.m.), the following table shows that for 2013 the average potential monthly revenue for a 100 kWh fleet was \$135.01.

Table 9. 2013 Potential Revenue from Participating in the 5:00 – 9:00 p.m. Timeframe

Hour Ending	\$ per 100 kWh	30 Days \$
18:00	\$ 1.57	\$ 47.01
19:00	\$ 1.28	\$ 38.50
20:00	\$ 0.92	\$ 27.50
21:00	\$ 0.73	\$ 22.00
Total		\$ 135.01

Up through March 15, 2014, as shown below, the peak moved substantially, and also the price for frequency regulation services increased significantly.

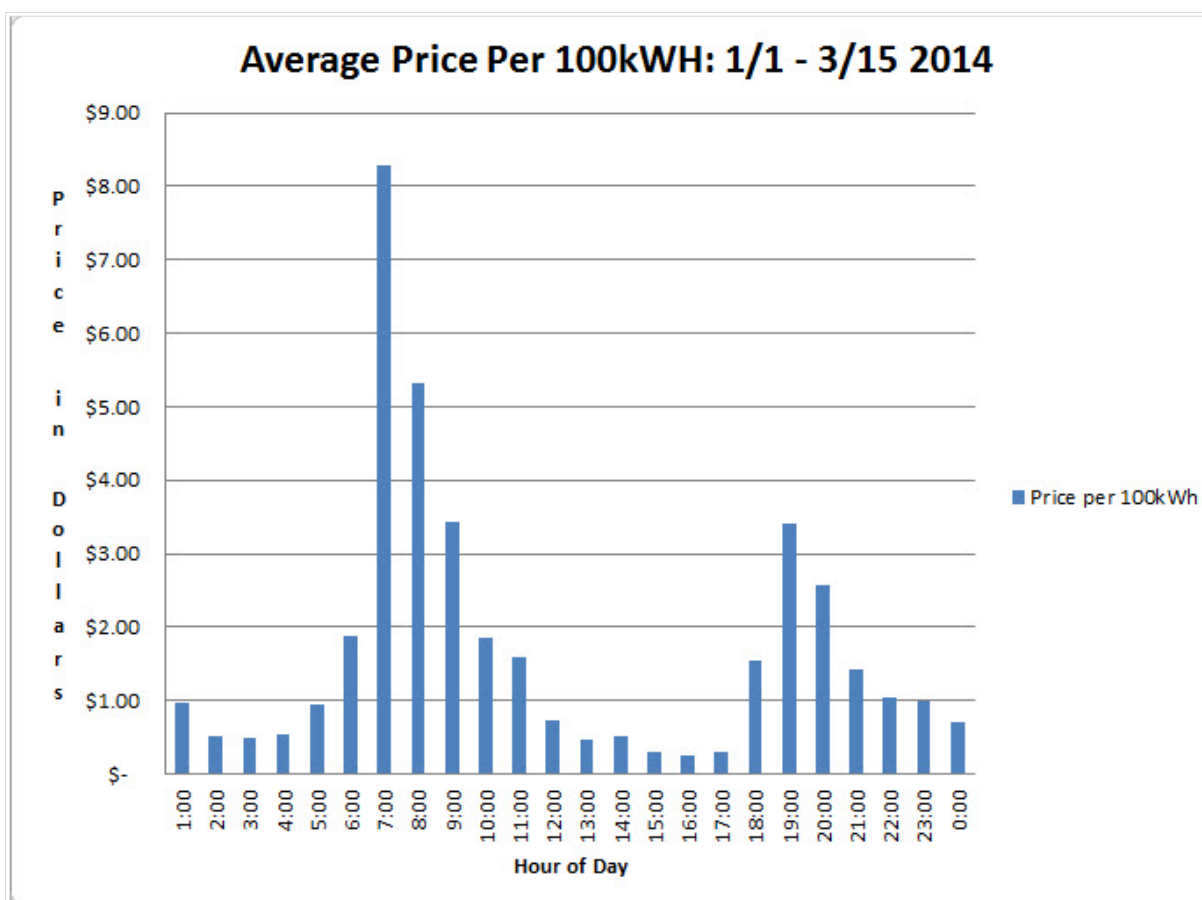


Figure 31. Average Frequency Regulation Price per Hour per 100 kWh in ERCOT for 2013

While the morning opportunity was not available for the project fleet due to their delivery schedules, the increase in price does have a positive effect on potential revenue, as shown below.

Table 10. 2014 Potential Revenue from Participating in the 5:00 p.m. – 9:00 p.m. Timeframe

Hour Ending	Price per 100 kWh	30 Days \$
18:00	\$ 1.55	\$ 46.53
19:00	\$ 3.40	\$ 101.97
20:00	\$ 2.58	\$ 77.27
21:00	\$ 1.42	\$ 42.51
	Total	\$ 268.29

The data shows that for 2014 the potential revenue was double that which was achievable in 2013. Even so, this was not sufficient to offset the costs of equipment, installation, and ongoing expenses. For example, all assets must go through a QSE to participate in market services, so there is a monthly price assessed by the QSE for maintaining the communications links to the asset and for representing them in the market (placing bids). The price for the QSE services for this project was \$2,000 per month.

In order to be profitable in the current market structure, the fleet would need to be of an approximate capacity size of 1 MWh, however that is very unlikely. Alternatively, if the fleet could be flexible in its scheduling, it could place bids in more opportune time periods, reducing the size of the fleet needed to participate successfully from 1 MWh to 400 kWh, as shown below.

Table 11. 2014 Potential Revenue from Participating in the 6:00 a.m. – 11:00 a.m. Timeframe

Hour Ending	Price per 100 kWh	30 Days \$
6:00	\$ 1.89	\$ 56.63
7:00	\$ 8.28	\$ 248.53
8:00	\$ 5.33	\$ 159.89
9:00	\$ 3.42	\$ 102.72
10:00	\$ 1.86	\$ 55.75
11:00	\$ 1.60	\$ 47.91
	Total	\$ 671.43

There are also the potential jackpot scenarios, catching an hour where the price of energy skyrockets due to grid congestion. While this effort was operational in 2014 there were at least 4 such hours where the cost per MWh ranged from \$284 to \$4,999. Those hours were normalized in the calculations to remove skewing of the data.

This project, as stated earlier, was conducted in a market where energy is relatively inexpensive and plentiful, and the market is structured for large, stationary generation resources. In other markets, such

as California where the price of electricity is significantly higher, the economics would be much more favorable.

General Conclusions and Key Lessons Learned from Discovery Across Texas

- Demonstration projects carried out on real operating systems have the advantage of high relevance, but it takes constant effort to involve stakeholders and garner support to overcome potential constraints
- Maintaining an active relationship with key market stakeholders who are the target users of the results is essential to project stability and effective outcomes
- Synchrophasor technology is here to stay, it has numerous demonstrated benefits, and it will soon become a very critical guide and asset in control rooms for grid operators
- Cyber security protection of grid systems is of rising concern and the prospects are good that the Security Fabric product demonstrated in this project will find market acceptance
- Battery storage has a number of challenges that must be addressed before the technology becomes mainstream in electric markets, including regulatory rules that permit the dual role of both generation and load, and stakeholders must implement creative approaches that enable them to maximize the benefit streams from multiple functionalities
- Interoperability among new devices and software in home energy management systems is a major market development inhibitor; radically low cost technologies coupled with market incentives are essential for DR to become economical and provide needed grid support
- Electric vehicles are potentially valuable grid resources but only special niche market players will participate in related AS markets in the near term
- Energy efficient, high technology communities matched with rooftop solar and electric vehicles can be a valuable grid resource with the advent of time-of-use pricing to reshape load, but without incentives the costs will continue to limit the scale
- Policy and technology applications for all-solar communities need to address the problem of harmonic distortion that appear to be common characteristics of solar communities

Most Likely Outcomes of the Discovery Across Texas Project

CCET engaged some 22 firms to carry out seven related but separable projects all devoted to helping address the challenge of wind integration in the ERCOT market. The projects were all carried out on live operating systems in the market place rather than isolated laboratory experiments. The projects were not developed in a vacuum based on an ideal optimal mix of technology options for wind integration. To the contrary, they leveraged the availability of technologies already in or entering the market such as synchrophasor systems, cyber security systems, the build out of the CREZ transmission system, smart meter deployments, consumer portals, wind test facilities and wind turbines, smart appliances and electric vehicles in new solar residential communities, and electric vehicle fleets.

Market stakeholders, some of whom were involved in this project, will take the results of this work to test new technologies and approaches as states like Texas expand levels of wind and solar penetration to historic proportions. ERCOT will continue overseeing and expanding the synchrophasor system, South Plains Electric Cooperative and Texas Tech University will further leverage the battery capabilities, REPs in the ERCOT market will learn from the consumer behavior responses to develop more enticing pricing programs as well as enhanced DR capabilities, Intel/McAfee will leverage the Security Fabric work to move the product into the growing world-wide market for cyber security protection, and large companies and potential aggregators will make use of the FRRS work in developing improved benefit streams for EV trucks and possibly even EV automobiles in more favorable U.S. markets.

Preface

The Center for the Commercialization of Electric Technologies (CCET) has prepared this Final Technical Report for its smart grid regional demonstration project (SGRDP) in response to specified Department of Energy (DOE) reporting requirements. The format of this document follows DOE's Guidelines dated June 17, 2011.

As the final technical report, this document builds on the information that was provided in previous Technology Performance Reports (TPRs), the last of which was delivered in April 2014. It covers the project scope and overview, and then provides individual chapters for each of the 7 project components:

- Synchrophasor monitoring, visualization and event reporting
- Synchrophasor security fabric
- Residential circuit monitoring
- Residential time-of-use pricing trials
- Residential demand response
- Distribution-level battery energy storage system
- Fast response regulation service with electric vehicles

Each chapter addresses the components in terms of anticipated benefits, stakeholder interactions, technical approach, performance results with metrics and benefits analyses, and conclusions with lessons learned and best practices.

1. INTRODUCTION

This document represents the final technical report for the CCET SGRDP entitled, “Technology Solutions for Wind Integration in Electric Reliability Council of Texas,” commonly known as Discovery Across Texas. The CCET project was partially funded by DOE under the cooperative agreement DE-OE0000194 and partially funded by various CCET team members and stakeholders.

For further information on this report or any aspect of the project components, the relevant contacts are:

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2. SCOPE

This section provides relevant information on the overall project, the team members, the major demonstration components and objectives, the project milestones and schedule, the basic approach for interoperability and cyber security, and a list of acronyms. The subsequent sections 3 through 9 provide details on the planning, execution, and results of the seven project component demonstrations.

2.1 CCET

The CCET is a consortium of electric utilities, retail electric providers (REPs), and technology companies dedicated to developing new capabilities that will improve grid operations in Texas, and primarily for the Electric Reliability Council of Texas (ERCOT) region which covers about 85% of the state. CCET has performed a number of demonstration projects during its eight year history, the largest of which is this DOE project which involves demonstrations of innovative technologies to address challenges in seven different technology areas.

2.2 Overall Project Description

This project represented a multi-faceted synergistic approach to managing fluctuations in renewables, and more specifically wind power, in the large ERCOT electric grid through better system monitoring capabilities, enhanced operator visualization, and improved load management. The project demonstrated seven different technologies, organized around three main focus areas:

- **Synchrophasor:** It demonstrated the use of synchrophasor technology to assist in better determining grid operating status and margins that avoid either conservative operating margins or grid instability when moving remote wind resources through the ERCOT transmission grid to consumers. By taking advantage of this technology, ERCOT grid operators are now better able to accommodate wind resource variability, unexpected transmission line outages, or instability developments.
- **Smart Meter Texas Portal:** It demonstrated how ERCOT operators, transmission and distribution service providers (TDSPs), and REPs can leverage existing smart grid capabilities, including the Smart Meter Texas (SMT) Portal, to call upon controllable home area network (HAN) loads as part of initiated demand response (DR) events via the advanced meter systems (AMS) of utilities.
- **Future Smart Grid Community:** It demonstrated the benefits and potential issues associated with smart grid communities of the future that combine distribution-level energy storage systems and homes equipped with smart meters, solar photovoltaic (PV), load-interruptible DR appliances, efficient building envelope standards, and electric vehicle (EV) charging to demonstrate consumer responsiveness to variable electric pricing incentives.

All together, this project helped to define a new business model for managing renewable resources using a variety of combined yet separable smart grid components. ERCOT and other market participants are better informed and equipped to utilize this project's collection of tools to meet reliability needs across the entire interconnection in response to renewable fluctuations and a variety of grid disturbances. Further, the project helped the ERCOT competitive market better understand consumer behavior in the developing world of new technology that provides insight for new consumer incentives and program offerings that better integrate renewables into the grid.

2.3 Team Members

This CCET project represented a collaborative effort on the part of four ERCOT region utilities, electric cooperatives, a large Texas REP, a university, an EV fleet owner, third-party vendors, and technology firms. Additionally, this project benefited from advice and guidance provided by the CCET Board of Directors, a Project Management (PM) Board, and ERCOT, all of which comprised key players in the Texas electric market.

To address the project goals, CCET employed technical teams to address three high-level demonstration areas: synchrophasors, future smart grid community, and the SMT Portal. The final high level team organizational structure is shown below.

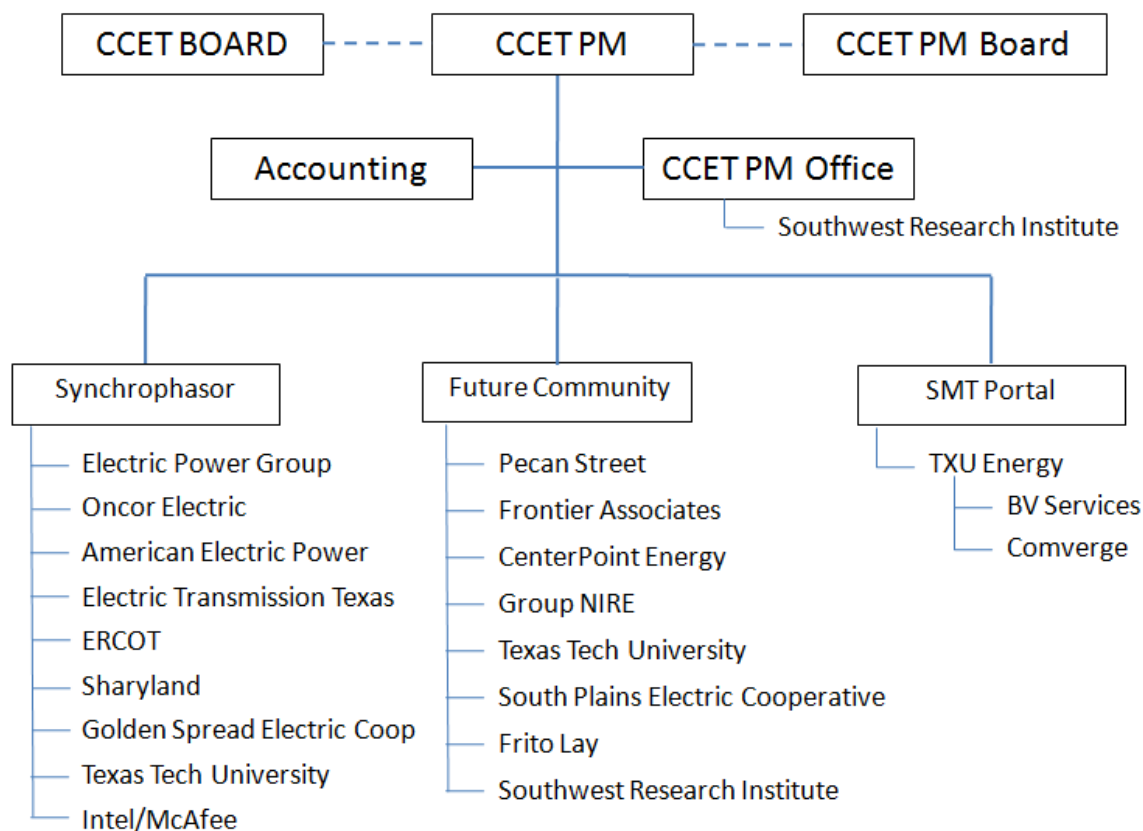


Figure 32. Final CCET Project Team Organization

The final team member descriptions are provided below, and organized by those that received DOE funds, those that provided cost share, and others that provided technical advice and oversight.

2.3.1 CCET

The CCET President served as the Project Manager (PM) and was fully responsible for the successful execution of this project. He is governed by a Board of Directors that approves all projects. The CCET PM had access to Board members for assistance and support, and reported monthly to the Board on the programmatic status of the project. The CCET PM also had access to a special PM Board formed by technical leaders of the Board companies, and met monthly with this group for advice and guidance on the conduct of this project.

2.3.2 DOE-Funded Partners

2.3.2.1 BV Services

For this project, BV Services assisted CCET and TXU Energy with the SMT Planning portion, which involved dual-path demand response services. BV Services planned and managed this part of the project as well as led technical discussions and workshops regarding solution configurations, program design, testing approaches, and impacts to customers. BV Services resources have been working with

TXU Energy and Comverge to support enhancements to the TXU Brighten® iThermostat program for the past couple of years.

2.3.2.2 Comverge

Comverge provides intelligent energy management solutions to utilities and commercial and industrial companies. Comverge worked with TXU Energy to develop its current Brighten® iThermostat program for residential and small commercial customers, which includes installing programmable thermostats and gateways, and providing software to manage the registration, installation, operation, and support of that program. For this project Comverge participated in the design, development, and testing of DR solutions.

2.3.2.3 Electric Power Group

Electric Power Group (EPG), a CCET member, served as the Team Lead for the Synchrophasor component. EPG has been involved in synchrophasor-related activities for many years in support of the Western and Eastern Interconnections. For this project, EPG provided software, including the Real Time Dynamics Monitoring System (RTDMS) and the Phasor Grid Dynamics Analyzer (PGDA), and professional and consulting services. They also have an enhanced phasor data concentrator (ePDC) which has been deployed as well. EPG is also closely involved in the development and implementation of ERCOT's synchrophasor monitoring system, and will continue to provide that support.

In early 2013, EPG formed an alliance with Intel/McAfee to investigate a new cyber security capability that greatly enhances the protection of data streams from PMUs and PDCs to the grid operator. As part of this project, EPG integrated that new capability into their suite of products and then demonstrated its ability to provide enhanced protection for the critical data streams.

2.3.2.4 Frontier Associates

Frontier Associates, a CCET member, has been providing data collection, analysis, and various assessments for key market players in Texas, including utilities and ERCOT. For this project, they were responsible for the overall data collection and analysis planning, the methodology for the pricing experiments, and analysis of electrical data collected by utilities on residential distribution service lines in Houston and Austin. They developed the metrics and benefits reporting plan for those analyses. They helped to design the consumer pricing experiments, and provided the majority of data analysis, economic value assessments, and other inputs to the reports. They also provided advice and guidance to other project members performing data collection and analysis.

2.3.2.5 Group NIRE

Group NIRE, a CCET member, is a clean energy development company that provides project development, finance and consulting services. It is currently developing wind projects and working with

several renewable energy manufacturers of original equipment and component to commercialize new products and technologies. Group NIRE is affiliated with the National Institute for Renewable Energy and with the National Wind Resource Center.

Although Group NIRE received some minimal DOE funding, it was also a major cost share contributor for this project, and provided facilities for field testing and validation of various wind turbines at the Reese Technology Center (RTC) in Lubbock, Texas. For this project, Group NIRE facilitated access to the wind turbines and managed the installation of a utility-grade battery energy storage system that was tested in various functional roles.

2.3.2.6 Pecan Street

Headquartered at The University of Texas at Austin, Pecan Street Inc. is a nonprofit research and development organization focused on developing and testing advanced technology, business models, and customer behavior surrounding advanced energy management systems. Pecan Street is focused on advancing understanding and solutions addressing utility system reliability, climate change, renewable energy integration, and customer needs and preferences. Its specific research expertise consists of creating, managing, protecting, analyzing, and responsibly sharing the highest quality original research data on how customers use electricity, natural gas, and water in their homes and businesses. For the CCET project, Pecan Street received DOE funds but was also a significant cost share partner. It was involved in facilitating new electricity pricing models, electric vehicle research, and sub-circuit level energy use data collection among select volunteer participants located in Austin's Mueller community. It also provided engineering analysis of power quality data at the point of utility connection as affected by rooftop solar PV.

2.3.2.7 Southwest Research Institute

Southwest Research Institute (SwRI), a CCET member, is a large independent, nonprofit organization located in San Antonio, Texas, that provides applied research and development for both government and commercial clients. It has 11 technical divisions that offer a wide range of technical expertise and services in such areas as chemistry, space science, nondestructive evaluation, automation, engine design, mechanical engineering, electronics, and automation and data systems.

For the CCET project, SwRI provided project management office (PMO) support including the preparation of required management and technical reports for CCET and DOE. SwRI also provided some early cyber security evaluations and led the effort to provide fast response regulation service (FRRS) to ERCOT using fleet EVs.

2.3.2.8 Texas Tech University

Texas Tech University (TTU) is a teaching and research institution located in Lubbock, Texas. It is home to the Wind and Science Engineering (WiSE) Research Center, which is focused on how to efficiently

harvest wind energy and mitigate wind-related damage. It is the only university in the nation that offers degree programs, from Bachelor to Ph.D., in wind energy related studies.

For the CCET project, TTU provided its exceptional knowledge, expertise and analytic skills to facilitate the integration of wind turbines located at the RTC with a utility-scale battery system along with subsequent analysis of battery functionality to support dynamic, variable energy sources like wind. TTU also worked with several local electric cooperatives to install a synchrophasor network in the Texas Panhandle portion of the Southwest Power Pool (SPP). This network served to augment the data from the ERCOT network for the purposes of theory and model development and validation. The network was additionally utilized as a test-bed for the synchrophasor cyber security demonstration, where TTU cybersecurity researchers provided an independent analysis of the security system.

2.3.3 Cost Share Only Partners

2.3.3.1 American Electric Power/Electric Transmission Texas

American Electric Power (AEP) is a Board member of CCET and has been actively involved in the deployment of synchrophasor components. AEP committed resources to this effort and worked collaboratively with ERCOT, Oncor, Sharyland Utilities, and Electric Power Group to accomplish the objectives of the synchrophasor component. AEP installed PMUs, provided phasor data to ERCOT, and determined the best utilization of phasor data for management of AEP stations and the ERCOT grid.

Electric Transmission Texas (ETT) is a joint venture between subsidiaries of American Electric Power and MidAmerican Energy Holdings Company. ETT owns and operates transmission facilities within the ERCOT region of Texas. For this project, ETT installed a number of PMUs and provided the phasor data to ERCOT.

2.3.3.2 CenterPoint Energy

CenterPoint Energy is the TDSP for the Houston area, and a Board member of CCET. CenterPoint completed its own smart grid investment grant that provided for the deployment of smart meters and a distribution automation system in their service territory. As part of this project, they installed a number of circuit monitoring devices with the intent of comparing load shape curves among two similar neighborhoods, one with high-efficiency homes and rooftop solar PV, and another without solar. They collected data from those devices, and provided engineering input to the reports on observations of correlations between power usage, load profiles, voltage levels voltage and current harmonic distortion and power factor for both communities. Also, by comparing similarly sized homes at the two neighboring subdivisions, analysis shows the extent to which there is a sustained decrease in power usage and reduction in peak demand from solar-equipped homes. CenterPoint also compared the voltage levels at the two neighborhoods.

2.3.3.3 Electric Reliability Council of Texas

The Electric Reliability Council of Texas (ERCOT) manages the flow of electric power to 24 million Texas customers, representing about 90 percent of the state's electric load. As the independent system operator for the region, ERCOT schedules power on an electric grid that connects more than 43,000 miles of high-voltage transmission lines and more than 550 generation units. ERCOT also performs financial settlement for the competitive wholesale bulk-power market and administers retail switching for 6.7 million premises in competitive choice areas. ERCOT is a membership-based 501(c) (4) nonprofit corporation, governed by a board of directors and subject to oversight by the Public Utility Commission of Texas and the Texas Legislature. ERCOT's members include consumers, cooperatives, generators, power marketers, retail electric providers, investor-owned electric utilities (transmission and distribution providers), and municipal-owned electric utilities.

The ERCOT serves as an advisor on the CCET Board of Directors, and provided synchrophasor analysis and event reporting for this CCET project. The staff produces a daily PMU performance report that is distributed to members of ERCOT, and the CCET synchrophasor team. They also introduced the capabilities of synchrophasor technology to organized ERCOT stakeholder committees that are important to the implementation of the technology for active grid management.

2.3.3.4 Oncor

Oncor, a CCET Board member, is a large transmission and distribution utility and also a key transmission avenue for wind energy from West Texas to other parts of the ERCOT grid for consumers. Oncor established a phasor network within the ERCOT grid, providing PMU installations and communications for streaming data to ERCOT. Oncor also supported the project's SMT planning efforts by conducting testing of various end use components.

2.3.3.5 Sharyland Utilities

Sharyland Utilities, a CCET member, was involved in the synchrophasor component of this project. Sharyland implemented PMUs at various locations, provided a phasor data stream to ERCOT, and determined the best use of phasor data for station management and the ERCOT grid.

2.3.3.6 TXU Energy

In 2002, the deregulated electricity market opened in Texas. At that time, TXU Energy began offering service as a certificated retail electric provider. One of their hallmark products is an intelligent thermostat and accompanying gateway that has been installed in thousands of consumer homes across Texas, and provides consumers with broadband access to the thermostats. For the CCET project, TXU Energy undertook a demand response program that involved sending load control signals through the Smart Meter Texas portal to the home thermostats and gateways. Those devices were also provisioned

to some consumer smart meters so that they received the signals via the AMS networks as well as via the residential broadband networks. TXU Energy is a Board member of CCET.

2.3.4 Other Partners Providing Unfunded Technical Assistance or Oversight

2.3.4.1 Frito Lay

Frito Lay has a distribution facility with 12 fleet EVs that make deliveries during the nighttime and morning hours up until noon, and are therefore idle during the remainder of the day. For the FRRS portion of the project, Frito-Lay provided the EV fleet and charging infrastructure to support this demonstration.

2.3.4.2 Golden Spread Electric Cooperative

The Golden Spread Electric Cooperative (GSEC) is a consumer-owned public utility that provides electric service for its rural distribution cooperative members. It serves consumers located in the Oklahoma Panhandle and an area covering 24 percent of Texas land area including the Panhandle, South Plains, and Edwards Plateau Regions. It provides the power and service for the South Plains Electric Cooperative which was involved in this project. For the CCET project, GSEC deployed a number of PMUs to provide comparative data on their area of Texas with respect to wind events.

2.3.4.3 Intel/McAfee

Intel Corporation is an American multinational semiconductor chip maker, and three years ago they acquired McAfee, a manufacturer of computer security technology. Together they have undertaken to integrate a wide range of security related products that tap into the security embedded in Intel hardware. For this project they provided extensive security expertise and security applications to support the Security Fabric demonstration which demonstrated significant cyber security protection to the synchrophasor technologies. Intel is a CCET member.

2.3.4.4 South Plains Electric Cooperative

The South Plains Electric Cooperative (SPEC) is a local, consumer-owned electric utility headquartered in Lubbock, Texas and serving the South Plains of West Texas. SPEC owns and operates the battery energy storage system purchased by the project and located at the RTC. SPEC is a CCET member.

2.3.5 CCET Advisory Boards

2.3.5.1 CCET Board of Directors

The CCET Board of Directors provided oversight of the project and assistance as needed. The Principal Investigator for the project is also the President and Chief Operating Officer of CCET and reports to the Board.

2.3.5.2 CCET PM Board

For this project, CCET created a PM Board from member CCET companies with specialized technical expertise in the project focus areas. The members of this Board provided unbiased, objective evaluations and assessments of all technical aspects of the project. This Board met monthly with the Principal Investigator and his team to review the technical project status, and provided advice and assistance on the direction of the project. Additionally, when the need arose, this Board convened to address specific technical issues or risks and provide early resolutions.

2.4 Project Overview

Texas currently has the largest installed wind generation capacity in the United States. With a growing electric load, the region served by the ERCOT grid reached an overall peak load of 68,305 megawatts (MW) in August 2011, and more recently reached a new winter record at 57,277 MW in January 2014. Increasing energy demand coupled with legislated renewable portfolio standards has focused efforts on inclusion of renewables and especially wind since 1999. As of October 2014, ERCOT had an installed wind capacity of 11,493 MW (see figure below) which provides about 10% of the ERCOT electricity needs.² In November of 2014, a new ERCOT wind record was set at 10,301 MW which represented nearly 33% of the load at that time.³ With the completion of new transmission lines into West Texas and the Panhandle area of Texas, it is estimated that the capacity of wind generation can grow to 18,500 MW, and a significant number of interconnection applications are signed and on file with ERCOT for wind farms that would take advantage of this new transmission capability (shown as Cumulative Planned in the figure below). Further, wind farms also are being added in the coastal areas of South Texas, presenting somewhat different power profiles to system operators than the wind from West Texas. All wind farms are distant from the main load centers, and present grid management challenges for ERCOT's system operators in accommodating both grid reliability and generation dispatch related to wind variability.

² ERCOT Monthly Operational Overview at http://www.ercot.com/content/committees/board/keydocs/2014/ERCOT_Monthly_Operational_Overview_201410.pdf

³ ERCOT Quick Facts at http://www.ercot.com/content/news/presentations/2014/ERCOT_Quick_Facts_112014.pdf

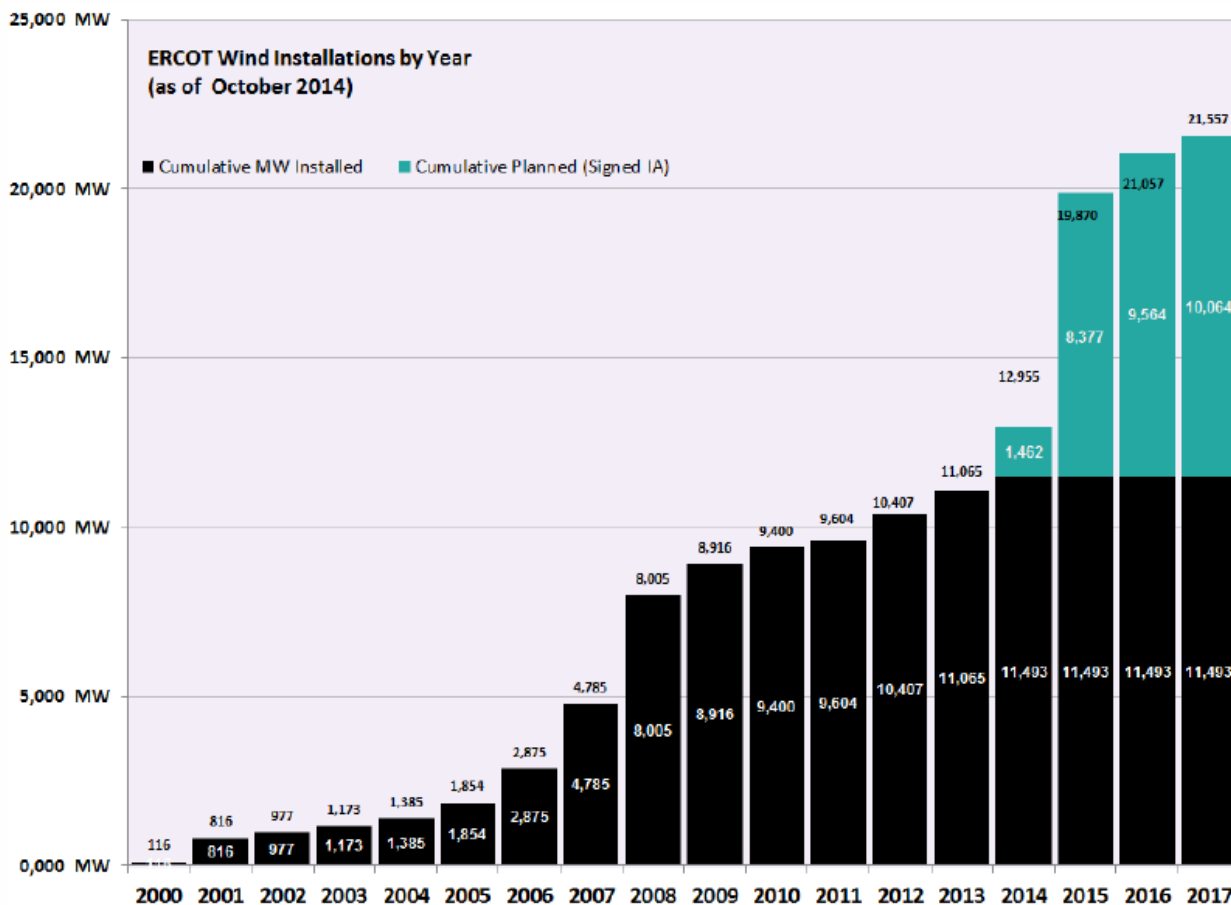


Figure 33. ERCOT Wind Installations by Year

The challenges to the Texas electric grid posed by this additional growth in wind generation underscores the need to establish new response mechanisms to wind variability. This project represented a multi-faceted synergistic approach to managing fluctuations in wind power in the large ERCOT transmission grid through better system monitoring capabilities, enhanced operator visualization, improved load management, and consumer response to pricing incentives.

2.4.1 Project Location

This project was primarily performed in the ERCOT region, although the utility-scale battery and some additional synchrophasor efforts were performed in the Northwest area of Texas that is actually in the Southwest Power Pool (SPP) region. These regions are shown in the figure below.

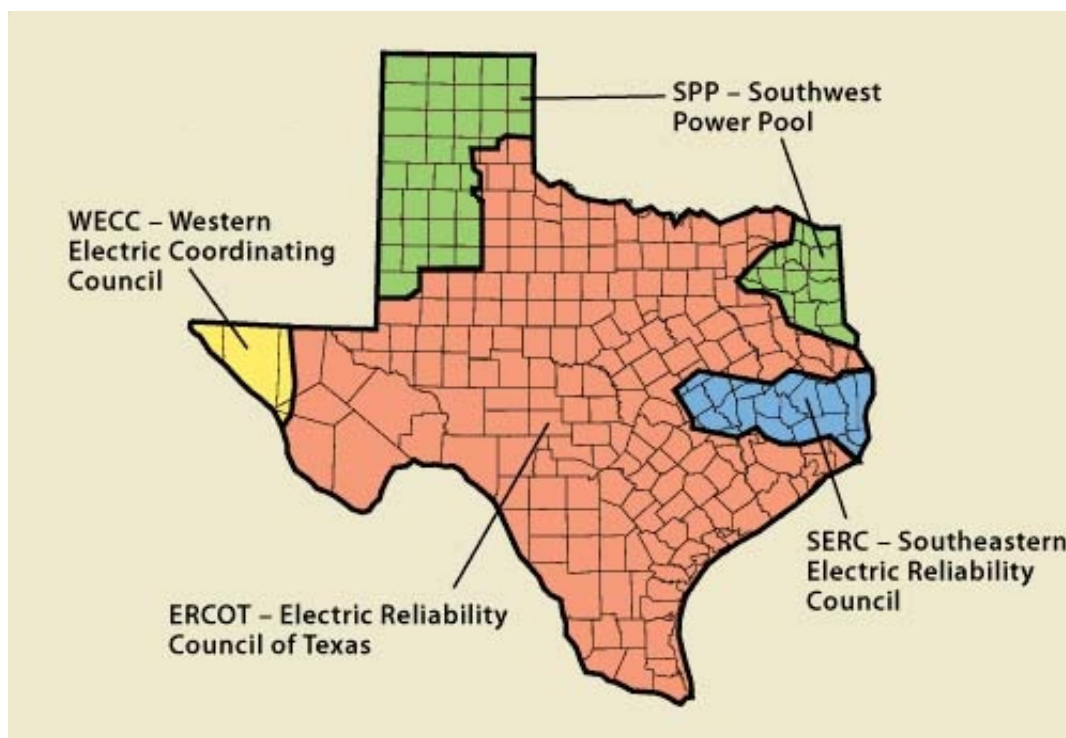


Figure 34. ERCOT and SPP Regions of Texas

Texas is unique among other regions with its high penetration of remote wind generation, one interconnection (ERCOT) entirely contained in the state, and one Independent System Operator (ISO), ERCOT. Texas includes a deregulated energy market for the large investor-owned segment, with TDSPs providing regulated wires and meter services to customers, and competitive REPs servicing and billing these customers. Municipal utilities and electric cooperatives which typically operate in a regulated market are invited to participate in the deregulated market. Texas not only leads the nation in wind generation plans, but also in the implementation of advanced metering systems. The most recent tally for the ERCOT region as of September 2014 ⁴ indicates at least 6,849,921 smart meters have been deployed, and there are plans for installation of even more smart meters in future years.

Another unique aspect of Texas is the availability of the web-based Smart Meter Texas (SMT) Portal which hosts the 15-minute incremental meter data from many of those smart meters and is used for settlement purposes as well as permitting customers to view and share their current and historical electrical usage information, and to bond home area network devices to their smart meter in support of demand response efforts.

⁴ AMWG Monthly Market Report at http://www.ercot.com/content/meetings/amwg/keydocs/2014/1022/AMWG_Monthly_Market_Reports_and_Formats_10_16_14_v1_0.ppt

2.4.2 Original Plan for Achieving Project Objectives

The purpose of this project was to demonstrate that Texas, and the rest of the nation, can manage increasing levels of wind power by accomplishing various objectives intended to deliver wide-area visualization through smart grid technologies applied at the system operator level; related operator control through DR as a resource utilizing AMS capability; wind variability control through energy storage; and application of DR, distributed generation (DG), and load reshaping at smart-grid equipped residential communities.

2.4.2.1 Demonstrating Wide-Area Grid Monitoring and Visualization using Synchrophasors

The Synchrophasor Team originally comprised ERCOT, TDSPs, and Electric Power Group (EPG). This group, through a prior CCET project, had already recognized the need to monitor the grid variability due to wind resources and had deployed phasor measurement units (PMUs) at three locations along with an initial version of visualization software at ERCOT. The objective of this aspect of the project was to expand the synchrophasor capability to at least 16 locations. The overall role was to stream data from the PMUs to the respective TDSPs who would then furnish it to the grid operator.

Through the prior CCET project and with DOE's support, ERCOT had a first generation software package for data management, visualization, and post-event analysis, and the plan was to improve the capabilities of this software. The objective was to potentially demonstrate the benefits of synchrophasor monitoring and analysis in effectively managing power reliability associated with variable wind power resources. An ultimate objective was to improve the software, the data streaming and processing capabilities, and the communications network to a level that would facilitate placement in the ERCOT control room as an ancillary tool for grid operators. All of this would require a collaborative effort to implement and maintain the supporting infrastructure, develop effective monitoring metrics, and identify, design, test, and deploy tools and procedures to improve overall grid operations and control based on the synchrophasor data system.

2.4.2.2 Demonstrating Aspects of the Future Smart Grid Residential Community

The Future Community Team originally comprised several entities in an area North of Houston. The goal of the project was to demonstrate the benefits of equipping an existing, developing solar community with a variety of smart grid technologies, including home area network (HAN) devices, home energy management (HEM) systems, residential storage systems, and plug-in electric vehicles (PEVs), and then perform various electric pricing experiments to measure consumer behavioral changes in response to variable pricing incentives. Additionally, the original concept included the monitoring of the solar community and a neighboring community at various feeders to define potential benefits and issues related to communities with heavy concentrations of solar and potentially PEVs.

Further, the original project envisioned deployment and analysis of a large battery system in conjunction with a solar farm to support community assets such as a water treatment plant and potentially community PEV charging stations.

2.4.2.3 Demonstrating Demand Response by Leveraging the Smart Meter Texas Portal

The TDSPs had jointly begun to develop an SMT Portal that would ultimately host and provide access to the 15-minute incremental smart meter data by REPs and consumers. The original objective was to collaborate with the SMT Portal Team and execute a planning effort to define additional functionality for enhancing the SMT Portal to support the project objectives.

2.4.3 Revised Plan for Achieving Project Objectives

In late 2011, due primarily to significant cost share issues, CCET was forced to cease all project activities until such time as it was able to restructure the project with several new partners. The changes in team composition, and additional activities that have become part of the project plan, are described below.

2.4.3.1 Synchrophasor

For the synchrophasor part of the project, the team was expanded to include Texas Tech University (TTU) which was interested in building its own synchrophasor network in conjunction with wind turbines located on a distribution network at the nearby Reese Technology Center. TTU is located in Lubbock, Texas, in the northern Texas panhandle area, which is part of the Southwest Power Pool (SPP) rather than ERCOT, so this provided an opportunity to investigate reliability in another area of Texas. Additionally, Golden Spread Electric Cooperative (GSEC) which provides generation and transmission resources for a number of rural electric cooperatives joined the team to provide additional locations for siting PMUs.

Furthermore, the synchrophasor efforts were expanded to include a demonstration of unique cyber security protection capabilities for synchrophasor assets by Intel/McAfee. The objective was to implement a parallel synchrophasor network in Lubbock, Texas, and have TTU perform independent cyber security testing of these new protection schemes.

2.4.3.2 Smart Grid Residential Community

The restructured project leveraged a solar community equipped with smart grid technologies in Austin, Texas, that provided a test bed similar to that originally planned in Houston. Pecan Street, which also has an SGRDP, and manages several related efforts in this community, was able to support the planned pricing experiments designed by team member Frontier Associates. The intent of conducting consumer pricing experiments remained the same, and this team planned, recruited participants, executed the experiments, and collected and organized the data from ongoing pricing trials supplied to Frontier Associates for analysis of consumer response to pricing experiments.

The utility-scale battery which was originally envisioned in line with a solar farm was situated in conjunction with adjacent wind turbines at the Reese Technology Center in Lubbock. The wind turbines and related projects are managed by Group NIRE, and these activities are situated on a local distribution network owned by South Plains Electric Cooperative (SPEC). They both have a close affiliation with TTU which performs analysis on the wind turbines and extended their scope to include analysis on the battery system.

2.4.3.3 SMT Portal

As the SMT Portal continued to be enhanced, TXU Energy, a REP, joined the project to ascertain the benefit of using the SMT Portal to improve its DR initiatives. Many of these DR programs already exist and typically operate by sending signals through residential broadband networks to various HAN devices that then curtail the devices or adjust settings to conserve energy use during critical times; however, for a variety of reasons, the broadband approach has seen limited success. The Team, therefore, focused on demonstrating the feasibility of improving those DR efforts by also transmitting DR signals through the SMT Portal and across the TDSP AMS networks to the same residential HAN devices located behind the smart meters.

2.4.4 Project Timeline

This section includes high-level schedules of the CCET project activities performed by the various project components. The project included three phases, and each phase is represented by a separate figure.

The figure below shows the primary scheduled activities for Phase I which was conducted from January through May 2010.

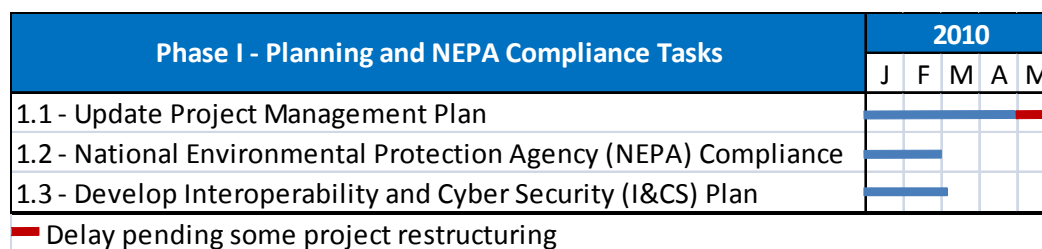


Figure 35. Primary Phase I Activities

This next schedule is for Phase II which began in June 2010 and concluded in October 2012.

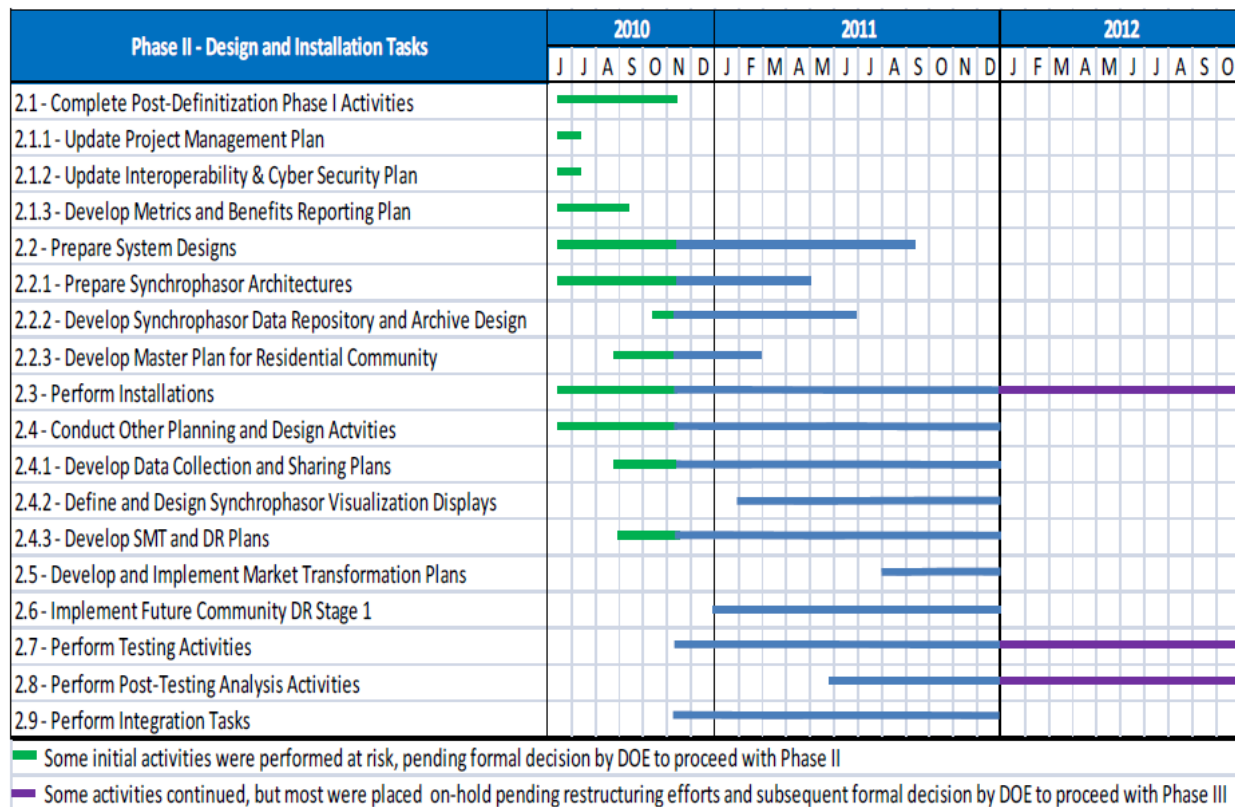


Figure 36. Primary Phase II Activities

The schedule shown below is for Phase III.

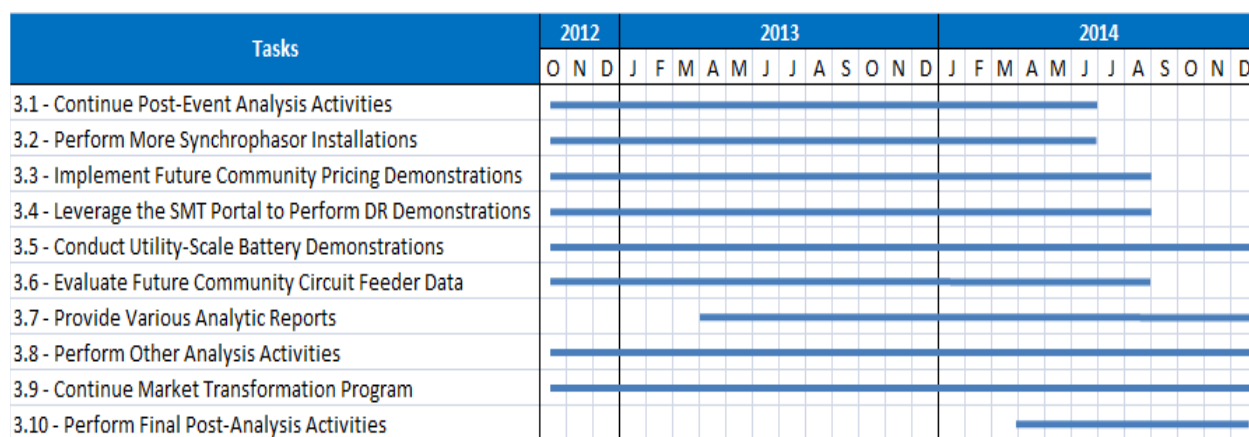


Figure 37. Primary Phase III Activities

2.4.5 Key Project Milestones

The key project milestones for the CCET project are listed in the table below. The milestones served as key verification dates for measuring progress, and they were typically characterized as deliverables to

provide a verification method for accomplishing the milestone on time. During project performance, CCET reported the milestone status as part of the required Monthly Progress Report.

The table identifies the respective project phase, milestone description and planned completion date.

Table 12. Key Project Milestones

Phase	Milestone Description	Planned Completion
I	Update the PMP	05/25/2010
I	Prepare NEPA Environmental Questionnaires	02/26/2010
I	Update I&CS Plan	03/08/2010
II	Update the PMP	11/26/2010
II	Update I&CS Plan	11/26/2010
II	Develop metrics and benefits reporting plan	01/28/2011, 05/16/2011
II	Complete system design documents	09/09/2011
II	Provide quarterly build metrics	Quarterly as of 3Q11
II	Create initial market transformation plan	12/16/2011
III	Update the PMP	12/14/2012
III	Update the I&CS Plan	01/15/2013
III	Update the metrics and benefits reporting plan	01/30/2013
III	Provide quarterly build metrics	Quarterly as of 4Q12
III	Prepare interim TPRs	09/30/2013, 04/30/2014
III	Prepare Final Report	12/30/2014

2.4.6 Key Decision Points

Decision points were used by the CCET Team and DOE to assess the technical status, results, funding requirements, programmatic needs, and relevant risk factors of the project. The success criteria at these decision points was used to assess achievement of specific goals which were typically at the end of each phase or at appropriate points in the project execution.

The success criteria shown below are objective and stated in terms of specific, measurable and repeatable data. Some of these success criteria pertained to desirable outcomes, results and observations from the demonstration efforts, and encompassed certain project aspects such as:

- Validation/confirmation/identification of scientific/engineering knowledge
- Cost savings expected over existing technologies
- Performance enhancements to existing technologies
- Reduction in health, safety and environmental risks
- Ease of installation, operation, and maintenance
- Decrease in capital, operating, and maintenance cost

Listed below are the decision points and success criteria for each one.

Decision Point 1 – Planning and NEPA Compliance

- Updated PMP completed and submitted to DOE
- NEPA Environmental Questionnaires completed and provided to DOE
- Interoperability & Cyber Security (I&CS) Plan completed and provided to DOE

Go/No-Go Decision Point 1 – End of Phase I

- Previous Decision Point 1 successfully accomplished
- Project efforts on schedule and within budget
- Plans for Phase II approved

Decision point 2 – Design and Installation

- System design documents completed and provided to DOE
- Market transformation plan completed and provided to DOE
- Quarterly build metrics provided to DOE

Go/No-Go Decision Point 2 – End of Phase II

- Previous Decision Point 2 successfully accomplished
- Project efforts on schedule and within budget
- Plans for Phase III approved

Decision Point 3 – Demonstration and Analysis

- Quarterly build metrics provided to DOE
- Interim TPRs provided to DOE
- Final report provided to DOE

2.5 Major Demonstration Components and Smart Grid Technologies and Systems

As summarized earlier in this report, the project began with three primary components: synchrophasor, future community, and SMT portal. As a result of the described restructuring effort that ultimately attracted new team partners, while retaining the same project goals and objectives, the project expanded its scope to cover seven technical demonstrations as part of these three components.

- Synchrophasor
 - Synchrophasor monitoring, visualization and event reporting
 - Synchrophasor Security Fabric

- Future Community
 - Residential circuit monitoring
 - Residential time-of-use pricing trial
 - Distribution-level battery energy storage system
 - Fast response regulation service with fleet electric vehicles
- SMT Portal
 - Residential demand response

For each of these technology components, there is a myriad of information to relate regarding aspects such as technologies and systems demonstrated, smart grid functions and applications, expected benefits, stakeholder interactions, project plans, data collection efforts, benefits analysis, results and conclusions. Rather than address the individual components in each of these areas, CCET has chosen to organize and present this information by technology component so that the reader gains a clear understanding of each component as a project, from inception through planning and demonstration execution to results and conclusions. The individual components are therefore contained in separate sections, as outlined below.

[Section 3.0](#) – Synchrophasor monitoring, visualization and event reporting

[Section 4.0](#) – Synchrophasor Security Fabric

[Section 5.0](#) – Residential circuit monitoring

[Section 6.0](#) – Residential time-of-use pricing trial

[Section 7.0](#) – Residential demand response

[Section 8.0](#) – Distribution-level battery energy storage system

[Section 9.0](#) – Fast response regulation service with fleet electric vehicles

2.6 Basic Interoperability and Cyber Security Approach

One of the key requirements for the DOE SGRDPs was to ensure that interoperability and cyber security were addressed. For this project, CCET chose a cross-cutting approach whereby interoperability and cyber security efforts were integrated into each technical component of the project. The technical approaches for performing these cross-cutting efforts are outlined in the next two sub-sections.

2.6.1 Interoperability

The CCET project members took measures to ensure that technologies in this project complied with documented standards for interoperability. During the project, the demonstrated technologies were exercised and evaluated to ensure they interoperated with both the existing electrical infrastructure as well as anticipated grid enhancements.

- **Synchrophasor network:** The teams chose devices from proven commercial vendors that had been in field use for years and had been validated by their vendors as satisfying interoperable

standards. Although communication issues were defined, these were related more to operating frequencies, environment and atmospheric rather than interoperability.

- **Security Fabric:** This suite of capabilities incorporates integrated devices and functionality that are commercially available individually. Interoperability testing among the suite of devices was performed by Intel/McAfee and the vendors proposing products for this suite.
- **Utility-scale BESS:** For the BESS, the vendor, the local cooperative SPEC, and TTU were responsible for any necessary interoperability testing. The BESS vendor performed laboratory testing and replicated that testing when the BESS was fielded. Since the BESS is a self-contained unit, using proven and reliable technologies, no issues were anticipated and none were revealed. There were still potential data transmission, storage and data visualization issues, but none of these were expected to be related to interoperability and field testing verified this.
- **Residential home area networks (HANs):** The HAN devices were chosen by the partners based partly on compliance with standards and partly on their technological capabilities. One set of these HAN devices selected for the pricing program was rigorously tested as part of its own demonstration project.
- **SMT Portal and DR:** Another set of HAN devices that were used for DR underwent both laboratory and field testing to demonstrate interoperability with both smart meters and broadband networks. The testing did in fact reveal that some older fielded devices actually had issues with interoperability and these findings were documented and revealed to the respective REP. To continue the DR program, the REP and CCET chose to leverage newly fielded devices that were guaranteed by the vendor to be compliant with interoperability standards. These devices also underwent both laboratory and field testing to validate their compliance.
- **Distribution-level monitoring systems:** The TDSP implementing the neighborhood circuit monitoring effort was responsible for performing all necessary interoperability testing.
- **FRRS:** The team followed established ERCOT standards implemented in similar efforts and verified through a certification process.

2.6.2 Cyber Security

The CCET project members recognized the importance of cyber security assessments as part of the project, and they were committed to ensuring adequate protection of all significant project components. However, many of the cyber security activities identified in NIST IR 7628 were either being performed outside the project scope, or the project activities had not yet matured to the point where cyber security assessments had been conducted.

As part of this project, CCET has implemented a compliant cyber security strategy to ensure protection of several key components:

- **Synchrophasor network:** The two largest Texas TDSPs and ERCOT implemented a network with PMUs and PDCs. This network employs secure communications, and the visualization results provide insights to grid operators, but the data and systems are not yet being used for grid control. When the latter occurs, the TDSPs will be responsible for expanding their current cyber

security audits to include their portions of the synchrophasor network. Additionally, CCET demonstrated a new data protection system (Security Fabric) at TTU that addresses the seven tenets of NIST IR 7628 and is designed specifically to monitor and deny intrusions between the TDSP PDC and the Independent System Operator's ePDC/RTDMS.

- **Security Fabric:** This suite of capabilities provides cyber security solutions, and both Intel and McAfee are national leaders in the security realm. Although they tested the protection capabilities of this suite, it was also independently tested by TTU as part of this project.
- **Utility-scale BESS:** During and after installation, the BESS underwent field testing to ensure that it satisfied all functional specifications, and was capable of performing all defined functions. As part of this effort, the BESS vendor performed cyber security assessments.
- **Residential home area networks (HANs):** This project leveraged HAN assets of another DOE SGDRP to evaluate the effect of various consumer pricing models. That other project was responsible for all cyber security assessments on their HAN devices and PEVs.
- **SMT Portal and DR:** This project included DR events utilizing HAN assets that belonged to a REP in Texas. These experiments leveraged the control capabilities of the SMT Portal to distribute pricing signals over TDSP networks to HAN devices located behind residential smart meters. The TDSPs are responsible for ensuring adequate cyber security protection on their meter networks, and the HAN device vendors and the REP ensured that the specific devices were compliant with all cyber security standards. Also, the SMT portal underwent a series of independent cyber security assessments and all findings were resolved.
- **Distribution-level monitoring systems:** One TDSP deployed monitoring systems on a distribution network to provide comparative analysis of electrical parameters in two different residential communities, one with solar and one without solar. That TDSP was responsible for ensuring adequate cyber security protection on that monitoring network. The distribution level monitoring was complemented by monitoring and analysis at the residential point of connection on a sampling of homes, performed by Pecan Street Inc.
- **FRRS:** This effort utilized data capture and reporting mechanisms that are similar to those in use by other market participants, and all of this was verified during the certification process.

3. SYNCHROPHASOR MONITORING, VISUALIZATION AND EVENT REPORTING

3.1 Introduction

Through a prior CCET project, team members had already recognized the need to monitor the grid variability due to wind resources, and they had deployed phasor measurement units (PMUs) at three locations along with an initial version of visualization software at ERCOT. The objective of this aspect of the project was to expand the synchrophasor capability to at least 16 locations. The overall role was to stream data from the PMUs to the respective TDSPs who would then furnish it to ERCOT, the grid operator.

Another objective was to demonstrate the benefits of synchrophasor monitoring and analysis in effectively managing power reliability associated with variable wind power resources. This was to be accomplished by improving the visualization software, the data streaming and processing capabilities, and the communications network to a level that would facilitate placement in the ERCOT control room as an ancillary tool for grid operators. All of this required a collaborative effort to implement and maintain the supporting infrastructure, develop effective monitoring metrics, and identify, design, test, and deploy tools and procedures to improve overall grid operations and control based on the synchrophasor data system.

This demonstration effort has enabled ERCOT to better manage the transmission grid to accommodate the very large quantities of wind generation that are coming onto the grid in remote locations, and to detect and identify potential undesirable conditions on the grid and enable ERCOT operators to make adjustments to resolve the conditions. These activities directly related to improving the overall operating reliability of the grid, and provided economic value by increasing the amount of wind that can be successfully integrated with the transmission grid.

This effort demonstrated a wide array of benefits related to the deployment of PMUs, including:

- A method for establishing and maintaining a reliable synchrophasor network to provide real-time dynamic information (i.e. voltage, frequency, location and time) on large-scale wind resources and their impact on the transmission grid, including assessing the optimal number and location of PMUs for effective wind integration, thereby advancing transmission operations management practices.
- The use of synchrophasor measurements to identify precursor conditions to undesirable grid performance and behavior, or to grid interruptions which, in turn, can lead to analysis, investigation, and ultimately changes in operating procedures or actions to facilitate integration

of intermittent resources, hence improving grid reliability when employing large amounts of variable renewable energy sources.

- Lessons learned from wind dynamics monitoring and management that can be replicated and transferred to other parts of the U.S., thereby promoting the use of variable renewable energy sources.

The PMUs, and the associated streaming, monitoring, and analysis capabilities, provide measurements that can be used for recalibrating engineering models, including voltages, transmission line loadings and angle measurements, small signal stability monitoring, and damping. They also provide measurements for identifying precursors to grid interruptions, such as low voltage margins, poor system damping, low frequency oscillations, and transmission system angles. Lessons learned that can be transferred from this project component's efforts include data adequacy, data quality, reliability and robustness for real time grid operations, wind integration, and reliability management.

The key project activities included:

- Establishing a robust synchrophasor network
- Validating and managing the synchrophasor data
- Performing analysis and assessment of grid performance based on synchrophasor data
- Reviewing and updating policies and procedures related to managing and maintaining the synchrophasor network and data storage archives.
- Supporting an ERCOT stakeholder process which is in the process of defining protocol changes needed to integrate synchrophasor monitoring technology into the control room.

At the inception of this project, the ERCOT network was being monitored by PMUs at three locations (one in northwest Texas, and two in southern Texas). Synchrophasor data was being captured and streamed from those devices to an enhanced phasor data concentrator (ePDC) at ERCOT. The data was then synchronized based on timestamps and displayed via the EPG Real Time Dynamics Monitoring System (RTDMS), which was running in the test environment at ERCOT. As part of this demonstration project, three TDSPs (Oncor Electric Delivery, American Electric Power, and Sharyland Utilities) committed to install PMUs at 13 additional locations, and to stream their synchrophasor data to ERCOT for monitoring and display in the RTDMS. After recognizing the true value proposition of this early warning network of devices, the TDSPs began adding PMUs at even more locations on their networks. As of November 2014, the TDSPs had installed and are providing streaming synchrophasor data from 76 PMUs at 35 locations to the ERCOT RTDMS. These efforts also attracted the attention of other utilities across Texas who began planning for deployment of additional PMUs. Currently, ERCOT is anticipating receiving data from nearly 90 PMUs at 46 different locations by early 2015. This network of PMUs currently provides monitoring of about 78% of ERCOT's regional grid footprint, and this will increase to about 85% coverage in 2015.

In addition to the ERCOT synchrophasor network, another objective of the project was to acquire data on the use of PMUs in tandem with wind turbines, and to evaluate their benefit to distribution grid

operations. For this aspect of the project, TTU worked with several electric cooperatives to deploy a synchrophasor network in the Texas Panhandle portion of SPP. This included devices that are directly measuring power from wind turbines at the RTC, not only providing data on the performance of the wind turbines, but also supporting a distribution grid, both of which were different from the ERCOT efforts. Additionally, this network provided a unique value given its location. As this network is electrically isolated from ERCOT, but located closely to the ERCOT transmission grids, it is likely that any events which occur in both systems can be attributed to environmental factors in the region such as high temperature or high winds. Additionally, this region is effectively a peninsula with limited connectivity to the rest of the SPP and the Eastern Interconnect. These factors made this synchrophasor network an ideal complement to the ERCOT system, providing an independent dataset for the testing of theories and the development of new engineering models which will aid future wind energy efforts.

3.2 Anticipated Benefits

Broadly, the expected benefits of the CCET demonstration project are a more reliable electric grid that can facilitate effective management of and responses to increased wind resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help grid operators better manage and respond to potential grid issues. Specific smart grid benefits that are supported by the synchrophasor component include:

- Provides transmission grid planners and operators with a high-resolution “picture” of conditions throughout the grid
- Furnishes critical parameters (voltage, current and frequency) that represent the overall health of the power system at much higher sampling rates
- Delivers improved power system monitoring and visualization to aid power system operators’ situational awareness and help them forestall grid collapse through better recognition and response to evolving grid events
- Assists in validation and derivation of system parameters used in power-system models and analytical tools to design and operate a more reliable grid
- Enhances grid throughput and utilization of existing grid assets
- Provides faster and improved forensic analysis following a disturbance that impacts the power system
- Provides necessary experience with synchrophasor systems to control room operators through training sessions and support of a stakeholder review process at ERCOT

3.3 Interactions with Key Stakeholders

The CCET synchrophasor team includes members from several of the largest utilities in Texas, and ERCOT, the grid operator for more than 85% of Texas. It also includes key members from the SPP area of Texas, including TTU and GSEC. It further includes one of the key application developers for the synchrophasor domain, EPG. Finally, team members are working closely with an ERCOT task force which

is building stakeholder consensus for control room use of the technology as a tool for improved grid management.

3.4 Technical Approach

This section provides a detailed description of the technical approach, the questions to be addressed, and the methods used to answer them. The first part, Project Plan, identifies the steps taken to demonstrate the technology. The second part, Data Collection and Benefits Analysis, defines the technical approach for conducting the benefits analysis.

3.4.1 Project Plan

This section covers the tasks that have been performed in support of the synchrophasor networks.

3.4.1.1 Key Tasks

For the Synchrophasor component, the Team performed the following tasks:

- Established a robust synchrophasor network
 - Designed and implemented a communications network capable of delivering synchrophasor data from each of the TDSPs to ERCOT in a timely and reliable manner
 - Designed and implemented a production grade computer environment to support the RTDMS wide-area monitoring and visualization system
 - Provided the capability for the participating TDSP operators to view the RTDMS displays remotely from their respective offices, thus providing the capability for both ERCOT and the TDSP operators to view the same screens and see the same information during grid events
 - Provided training to both ERCOT and TDSP operators in the use and interpretation of the RTDMS wide-area monitoring system
- Validated and managed the synchrophasor data
 - Evaluated the PMU data collection, reliability and robustness
 - Validated that all data sent by a TDSP is successfully and timely received by ERCOT
 - Validated that all data received by ERCOT is faithfully archived in the appropriate phasor data base
 - Evaluated the overall reliability of the PMU data communications network
 - Developed phasor data performance standards
 - Implemented and maintained a phasor performance monitoring system
- Performed analysis and assessment of events
 - Performed baseline angle separation and alarm limit analyses for the ERCOT grid
 - Performed post-event analysis and forensics on grid events and disturbances
 - Developed baseline performance measures for wind generation

- Assessed the impact of wind generation on system inertial and governor frequency response
- Assessed low voltage ride through performance of wind generation
- Detected, monitored and analyzed power system oscillations and the interaction of wind generation
- Implemented a means of validating model-based predictions of generator response to disturbances
- Developed, updated, and managed policies and procedures related to managing and maintaining the synchrophasor network and the data archives
 - Phasor data repository design and implementation requirements and data archiving policies
 - Data sharing policies (inside and outside ERCOT)
 - Phasor data management policies (e.g. PMU naming convention, change management)
 - PMU location selection principles and criteria
 - PMU use cases

3.4.1.2 ERCOT Synchrophasor Network

When the CCET effort began in January 2010, the ERCOT grid contained PMUs at three locations and an earlier version of the RTDMS visualization software. The plan under the DOE cooperative agreement was to increase this number of PMUs to 16 locations total, and to improve the RTDMS. Although not all directly related to the project, as of November 1, 2014, the ERCOT grid has 83 PMUs in service at 35 locations, and this number is expected to continue to increase to nearly 100 PMUs at 46 locations in the near future. ERCOT now recognizes the tremendous value that synchrophasors can offer its grid operators, so it is making plans to transition the data streams and internal monitoring software to production grade hardware so that operators can begin taking advantage of the alert features.

In recognition of the value that synchrophasors can provide to grid operations, ERCOT's Reliability Operations Subcommittee (ROS) created a Phasor Measurements Task Force (PMTF) in September 2013, comprised of representatives from TDSPs, Qualified Scheduling Entities (QSEs), Resource Entities, and ERCOT, and tasked with the following scope of work:

- How will synchrophasor data be used in ERCOT (real-time monitoring, unit model validation, etc.)? What are the current and planned uses by ERCOT? What are current and planned uses by TSPs and generators?
- What kinds of locations of PMUs are needed to meet this use(s)? How will these locations be determined?
- What are the data latency and quality requirements and other specifications, in order to meet the intended uses? Should these vary depending on the kind of PMU location?
- Who should install and own the PMUs? Who should be responsible for the communications from the PMUs?

- How should synchrophasor data be treated in terms of confidentiality?
- What reporting should be established for synchrophasor data?
- What data retention should be established for synchrophasor data?
- What are the CIP implications/requirements for the use of this data?

The PMTF's findings were presented to and accepted by the ROS in late October 2014, and the PMTF will be retired following its December 2014 meeting.

Regarding the network topology, the TDSPs have tried various communication configurations to transmit the PMU data to ERCOT. Those in well-established areas of the state have had fewer issues with communications, but it also has to do with placement of the PMUs. Initially, some TDSPs transmitted their streams directly from their PMUs to the ERCOT ePDC. However, since mid-2014, all of the TDSPs now transmit their data streams from their respective PDCs to the ERCOT ePDC.

3.4.1.3 TTU Synchrophasor Network

The synchrophasor network being deployed by TTU in the Texas Panhandle portion of SPP (shown in light blue in the figure below) began in late 2012 following the restructuring of the project. Currently, the system consists of five PMUs, an ePDC server located at the RTC and an RTDMS server located on the TTU campus. The first operational PMU is a National Instruments PMU located at the TTU campus. A second operational PMU is located in a Lyntegar Electric Cooperative-owned substation in Lynn County. Three additional PMUs are located at the RTC. Two of these PMUs are also functioning as revenue meters for two wind generation assets: the 1.67 MW Alstom wind turbine and three 300 kilowatt (kW) wind turbines at the SWiFT facility, while the third PMU is functioning as a revenue meter for the battery energy storage system (BESS) which is located near the SWiFT facility. The red stars in the figure below indicate the existing PMU locations.

An expansion of the TTU network to more locations is planned for early 2015. The yellow stars in the figure below indicate possible future locations. Three units are already planned for locations in Hale, Plains and Oldham counties, and this will significantly expand the geographical area of the network and provide PMUs located near significant wind or natural gas generation assets.

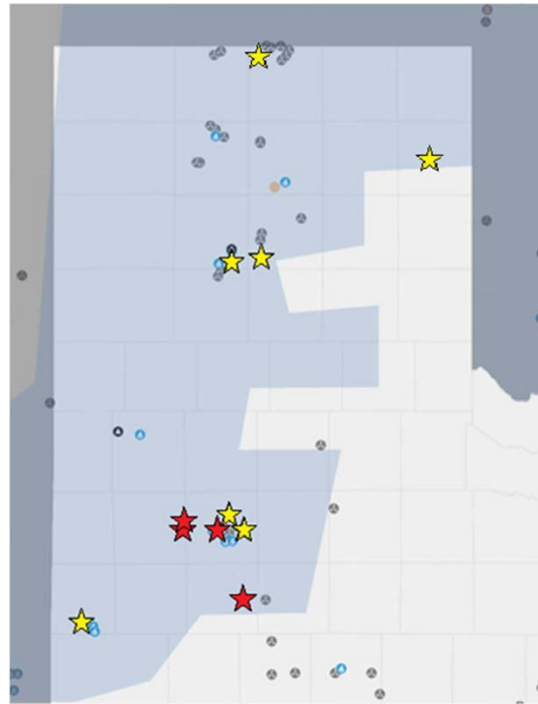


Figure 38. Current and Planned TTU PMU Locations

3.4.2 Data Collection and Benefits Analysis

To support the validation and analysis portions of the Synchrophasor component, the following data has been and will continue to be collected, retained, and analyzed:

- Synchrophasor data is retained in full resolution for the entire demonstration period. This data contains at least voltage, frequency, location, and time. Most of the PMUs sample at 30 samples per second while others record at 20 samples per second. This data has been used for the baselining analysis, together with State Estimator data recorded by the ERCOT Energy Management System (EMS). This data is also used to monitor the overall performance of the PMU network, including the communications network and the computer network infrastructure.
- Daily reports summarizing the overall performance of each of the PMUs reporting to the ERCOT ePDC are prepared and disseminated. These reports help to identify poorly performing PMUs, and identify communications and computer network delays which can interfere with reliable delivery and presentation of synchrophasor data to the ERCOT grid operators.
- ERCOT grid events, such as loss of generation or load, severe voltage dips, or sustained oscillations that are detected by the RTDMS wide-area monitoring and visualization system are also retained and analyzed. The events can be played back for review and operator training, and can be analyzed using an offline tool, the phasor grid dynamics analyzer (PGDA).

Additionally, the full synchrophasor data is scanned for unusual signatures, such as low-level oscillations that are characteristic of electromechanical interactions between generation and load, as well as

control-system driven oscillations. Synchrophasor data can also be used and compared to dynamic simulations to validate the performance of generator dynamic models with recorded performance.

The TTU portion of the synchrophasor network allows for positioning of PMUs at sites that have value for both wide area monitoring and for gathering data on individual generation or load assets. This ensures that the data set will be extremely useful for a wide variety of wind-related research. The TTU team is also identifying events such as generation trips and sub-harmonic oscillations and correlating those events with SPP wind energy contribution data, the operation of specific wind generation assets, or events which are also noted in ERCOT for which a cause has not been established.

3.5 Performance Results

This section covers the operation of smart grid technologies and systems, including the demonstration results; impact metrics and benefits analyses; and any stakeholder feedback received to date.

3.5.1 Operation of Smart Grid Technologies and Systems

This section primarily discusses the demonstration results for:

- PMU monitoring coverage
- Baselining study
- Data quality study
- Lessons learned from mitigating PMU data loss
- Event analyses of generation outages and wind interactions
- Requirements for control room and network production upgrades
- ERCOT and TDSP operator training
- Generator model validation
- Slow scan synchrophasors
- Synchrophasor use cases

3.5.1.1 PMU Monitoring Coverage

At the inception of this demonstration project, the ERCOT synchrophasor network consisted of PMUs at three locations and a single PDC installed and operated by American Electric Power (AEP), and a single PDC and Real Time Dynamics Monitoring System (RTDMS) wide area visualization system installed at ERCOT (installation was sponsored by CCET with DOE matching funding support). The demonstration for this project anticipated that the three participating TDSPs (Oncor Electric, AEP, and Sharyland Utilities) would install a total of 13 additional PMUs at 13 locations on their respective transmission networks, and provide the associated synchrophasor signals to ERCOT.

During the first two years of this project, the TDSPs identified additional locations where they could support the installation of PMUs. As of November 2014, the TDSPs have installed and are providing

streaming synchrophasor data from 76 PMUs⁵ at 35 locations across ERCOT as shown in the figure below. Although not all of the PMUs are directly funded by or formally part of this project, their data feeds certainly provide measured contributions to grid performance and event evaluations. Additionally, it is expected that other ERCOT TDSPs will soon be providing data from additional PMUs being installed in their substations to ERCOT. Currently, ERCOT is anticipating receiving data from nearly 90 PMUs (out of 100) at 46 different locations by the end of the year.

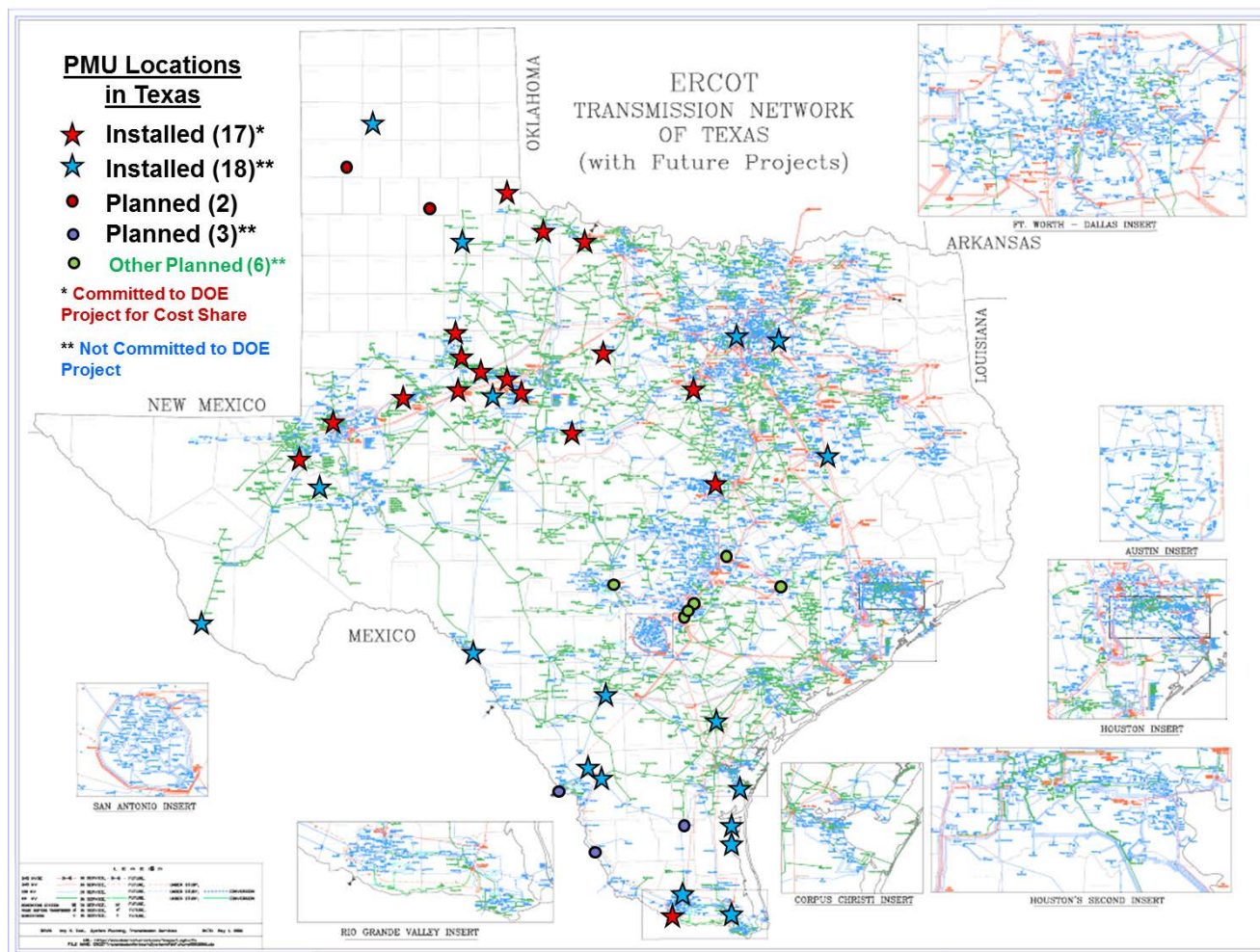


Figure 39. PMU Locations in Texas

During 2012 and the first part of 2013, the ERCOT ePDC received synchrophasor data streams from two TDSP-owned PDCs (Oncor and AEP). After adding new communications networks, Sharyland chose to stream its data directly from its two PMUs. This unusual data configuration worked quite well, and enabled the addition of new PMUs without the need to first install a local PDC at the TDSP. In mid-2014,

⁵ There are 83 PMUs currently installed, which includes multiple PMUs at some substations. Signals from 76 PMUs are being used for wide-area monitoring.

Sharyland redirected their two direct-feed PMUs into a Sharyland-owned PDC which supplies data (including three new PMU data streams) to the ERCOT ePDC.

At ERCOT, the ePDC and RTDMS had initially been running on servers in ERCOT's test environment while the operational support capabilities were evaluated. In early 2013, planning began for the installation of production grade servers and associated hardware. This migration was completed in September 2013.

At TTU, the ePDC and RTDMS systems have been installed and running since May 2013. Their synchrophasor network, which is in the SPP region (not ERCOT), currently consists of two stable PMU installs with high data availability and three additional installs which initially had very low data availability, but which are now performing reasonably well. The two high data availability PMUs are located on the TTU campus and at a Lyntegar-owned substation in Lynn County. The three lower availability PMUs are all located at the Reese Technology Center near the SWiFT wind turbine facilities. Two of the PMUs are connected over a 900 MHz encrypted Ethernet radio link to the BESS location, where the signals are moved onto a fiber optic communications circuit which connects to the Reese ePDC. The BESS PMU is also connected to this fiber system. An additional 3-6 PMUs are planned to be added to this network during 2015.

Initially, PMU radios linked to the Reese campus offices, where the ePDC is located. However, these links had low availability due to poor signal strength negatively impacted by line-of-sight problems. These two radio links were redirected to the battery installation in late 2013, where they are now tied to the fiber optic network, significantly improving the data availability. The battery is much closer to the PMUs and has excellent line-of-site to both locations. An additional challenge in achieving high data availability has been the processing/conversion of the data being extracted from the revenue meters (Lynn County substation and the three PMUs near the SWiFT facility at Reese). This data is converted from a proprietary communications protocol used by the revenue meter into a C37.118 protocol used by the ePDC via a custom-built data conversion algorithm. However, tuning the conversion algorithm to successfully identify and remove the additional end-of-field codes which the proprietary protocol occasionally generates has been challenging.

3.5.1.2 Baselining Study

In order to both validate the accuracy of the synchrophasor data being delivered to ERCOT, and to begin the process of establishing alert and alarm levels to inform the grid operators of emerging conditions on the system, a baselining study was initiated. The study evaluated both synchrophasor data and about 16,000 state estimator cases from the ERCOT energy management system (EMS) for the year 2012.

The synchrophasor data included all the PMUs that were installed and providing data to ERCOT (some were added during 2012, and thus covered only a portion of the year). The synchrophasor data was down-sampled from the nominal data rate of 30 samples per second to one sample per second to streamline the analysis.

The baselining analysis included a clustering study that identified PMU locations where the voltages and voltage angles moved in unison with nearby PMUs, as a means of identifying unique monitoring regions within the ERCOT grid. A total of six regions were identified, and similarly performing PMUs were thus clustered. The clustering study also compared the voltage magnitude and angles as measured by the synchrophasor network with the same metrics from the state estimator study results to validate the comparability of the two data monitoring systems.

The baselining analysis next analyzed, for the entire calendar year 2012, the ranges of voltage magnitude, voltage angles relative to a reference substation, and voltage angle differences (pairs) between PMUs in different regional clusters. This analysis was done for each of the anticipated 42 substations that are expected to ultimately have PMUs providing synchrophasor data to the ERCOT control center. The analysis also identified maximum and minimum values for each of the study metrics. Results from this baselining analysis were presented to the project team, and are attached to this report at the end of this section as Attachment 1.

Because the ERCOT grid was changing significantly between the start of 2012 and the end of 2013 with the addition of new 345 kV CREZ circuits that would dramatically alter the characteristics of the grid, it was determined that the experience-based operating limits that might be determined from this baselining study (based on the 2012 grid configuration) would not be reasonable limits based on the expanded ERCOT grid (including the new CREZ lines). Thus, it was determined to perform an update of the study based on the first six months of 2013.

As part of the 2012 baselining analysis, an analysis of the voltage magnitudes and voltage angles (relative to a reference substation) of PMUs which were geographically clustered was performed, to determine if representative PMUs from within clusters could be used individually to effectively monitor the overall ERCOT grid performance. The results of that analysis were included in the 2012 baselining analysis. For the 2013 baseline update analysis, it was recognized that new CREZ transmission lines had been added, which would change the electrical characteristics of the grid, and new PMUs (substations) had also been added. To verify the previously identified clusters, an updated cluster analysis was performed based on data from January-June 2013. The updated cluster analysis identified some minor changes in the makeup of the cluster groups. In general, the results corroborated the findings of the 2012 cluster analysis except for South 2, which did not track well with the other nearby substations. The updated cluster analysis report is attached as Attachment 2.

The first update study analyzed, for the period January-June 2013, the ranges of voltage magnitude, voltage angles relative to a reference substation, and voltage angle differences (pairs) between PMUs in different regional clusters, as did the 2012 Baselining study. This update study identified that, compared to the same time period in 2012, the ranges of voltage magnitudes, voltage angles relative to the reference substation, and voltage angle pairs for the first six months of 2013 had smaller deviations (minimum to maximum), and smaller angle spreads. This is consistent with the changes that would be expected when transmission lines are added between outlying locations and the load pockets to strengthen the grid. The first update study is attached to this report as Attachment 3.

During the first six months of 2013, nine new 345 kV transmission lines were completed as part of the CREZ project. Additionally, 19 new 345 kV transmission lines were completed between July 1 and December 19, 2013. As before, these lines were expected to change the electrical characteristics of the grid. To track those changes, data for the remainder of 2013 was collected and analyzed to monitor the changes in voltage magnitude and voltage angle ranges caused by the ongoing additions of CREZ transmission lines. Additionally, initial alarm setting values were determined for use in the ERCOT RTDMS system. This study update for the full twelve months of 2013, identified as Baseline Study Update #2, is attached as Attachment 4.

Because several CREZ transmission lines were completed late in 2013, and seven more lines were completed in the first quarter of 2014 (the last line was completed in March 2014), a third update of the baselining study was performed, based on data for February through July 2014. For this update, only voltage angles were analyzed, as the alarm limits based on voltage angles would have been most affected by the addition of new transmission lines. The conclusions from this third update study are:

1. The voltage angles for substations in the West Texas and Panhandle areas have tightened (smaller angles referenced to central Texas) significantly due to the addition during the second half of 2013 and January of 2014 of 22 new CREZ lines between these substations and the central area of Texas.
2. A re-distribution of power has occurred among the several transmission lines in the Valley area of Texas.
3. These voltage angle monthly medians are likely to change less in the future unless a significant amount of wind power is added in the western area and the Panhandle areas of Texas displacing generation in other areas such as the Valley area.
4. The new CREZ lines added in March 2014 are not expected to result in any further major changes in voltage angle monthly medians.

Recommended RTDMS Alarm Limits are included in this third update study. This final Baseline Study Update #3 is attached as Attachment 5.

3.5.1.3 Data Quality Study

In order for ERCOT to achieve a production quality phasor monitoring system that can be relied upon in real-time operations, three conditions must be met: 1) the data must be flowing reliably from the PMU to the grid operator's console, 2) the data must be valid, and 3) the data must be monitoring the critical locations (right places). In early 2013, an analysis of the overall synchrophasor data flow and data quality was initiated to address this first condition, and to identify improvements that might be needed to successfully achieve the first condition.

The goals of the study were to:

- Identify nodes in the phasor network that might be affecting data availability (dropout).

- Classify identified dropout issues by severity and frequency.
- Determine likely causes of data dropouts at identified locations.
- Propose solutions to help eliminate identified data availability problems.

Two of the TDSP's were providing their PMU data streams to ERCOT via TDSP-owned PDCs with associated data bases. To validate the flow of data, data extracts from each of the TDSP data bases were requested. However, for the specific dates selected for analysis, data from only one TDSP was available. It was decided to complete the initial study using data from a single TDSP, and then to repeat the analysis for the other TDSPs at a later date. While the initial study results concluded that all the sub-second synchrophasor data was reliably being delivered from the TDSP into the ERCOT ePDC and RTDMS databases, there were gaps in some of the second-average and minute-average data that was also being saved in the RTDMS database. An incompatibility between the data and the database storage routines was identified and corrected in early July. A review of the data base consistency has confirmed that this has corrected all of the identified issues from the initial study. The study was then extended to analyze the data streams from the other two TDSPs. A review of the data base consistency between each TDSP's local data base with the ERCOT database identified minor issues (primarily associated with local storage and data quality flagging). The Final Data Quality Study is attached as Attachment 6.

As part of their scope, the ERCOT PMTF updated and enhanced the ERCOT Communications Handbook to include a section on synchrophasors. The Data Quality validation process that was used for the aforementioned Data Quality Study has been incorporated into the ERCOT Communications Handbook, as a guide for ERCOT market participants and TDSPs who plan to add PMUs to their respective equipment. A copy of the Data Quality section of the ERCOT Communications Handbook is attached for reference as Attachment 7.

3.5.1.4 Lessons Learned from Mitigating PMU Data Loss

During the first two years of this project, the synchrophasor data volume has increased from 3 PMUs reporting at the outset to 83 PMUs reporting as of November 2014 (of which 76 are being used for operations monitoring). Since September 2011, daily reporting statistics have been tracked for each of the PMUs. As indicated in the summary plot below, which presents monthly average values through November 2014, early in the project when new PMUs were added, their availability and data quality was initially low but improved rapidly. As the participants have gained experience adding PMUs in their respective networks, the initial reporting performance of the newer PMUs has started higher and improved more rapidly. The Data Availability metric has been essentially 100% since identification and correction of a data base storage timing issue in early 2014—an issue identified as part of the Data Quality Study.

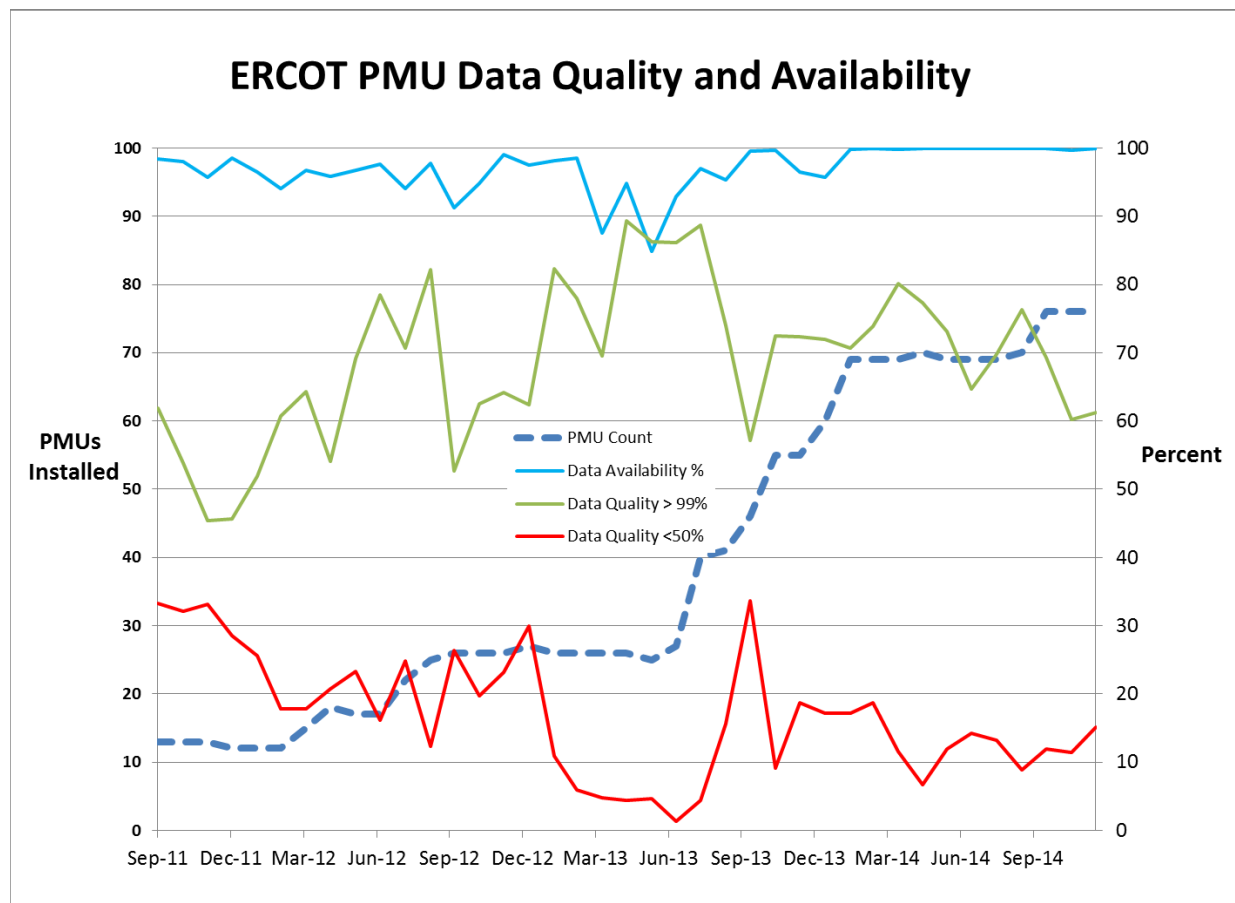


Figure 40. ERCOT PMU Data Quality and Availability

The most common challenges to achieving high overall PMU data quality have been securing a reliable communications channel between the PMU and the TDSP PDC, and getting a good time stamp signal into the PMU. As is characteristic for most utilities in North America, the ERCOT grid is represented by both legacy TDSP communications networks, and newer, more robust communications paths. Where PMUs were relying on the legacy communications networks, some tuning of the network was needed in order to improve the overall reliability of the PMU data stream. In three cases, it was not possible to achieve the necessary bandwidth and communications reliability to support the PMU data stream, so those PMUs were removed from the data network.

Sharyland initially attempted to utilize a wireless carrier for the first link of their PMU data stream from one of their substations. The wireless carrier disconnected the data stream after only a few days of operation, and refused to provide service to Sharyland—the carrier was not anticipating the volume of data represented by the PMU data stream. To complete the communications link to their substation, Sharyland had a hard-wire communications line installed.

Two PMU signals in the AEP PMU data stream are being provided voluntarily by their respective generation plant owners. Unfortunately, since providing PMU data is not high on the plant owners' priorities, these units are not being adequately supported, so the data quality from these PMUs is lower than desired, and the signals are not as useful as they might otherwise be. These and other minor local issues are the primary reasons for the metric of Data Quality <50% to be remaining up at about 10%. At this time, ERCOT has elected to continue utilizing these lower quality signals, which still provide some useful monitoring, rather than turning these signals off.

On one occasion, a network card on a co-located ERCOT-owned switching device, in the AEP facility where AEP's T-1 connection interconnected with the ERCOT wide area network, was causing random delays in transmission, causing the AEP PMU data stream to drop out for short periods of time. Diagnostic testing by both AEP and ERCOT validated that connections to their respective ports on the network switch were working properly, but testing across the switch was not performed. After several weeks of poor performance, an onsite technician removed and reseated the network card, duplicating the problem. The network card was replaced, and the intermittent data dropouts ceased. In this case, neither organization was sufficiently aware of the interface configuration to be able to test across the switch as part of their routine diagnostic testing.

3.5.1.5 Event Analysis of Generation Outages

The ERCOT RTDMS system automatically captures event files when significant generation is lost (e.g., loss of a unit 400 MW or greater). These event files are then analyzed to assess the overall performance of the grid, including frequency dip and initial inertial frequency recovery, to compare against the interconnection frequency bias, voltage dips (location and magnitude), and angle swings across the grid. These results are being accumulated and tabulated, together with grid configuration metrics such as total amount of generation on-line, mix of generation on-line, level of wind production, etc. Statistical analysis of these event metrics will be used to identify performance trends.

Additionally, on several occasions, minor sustained oscillations have been detected on the ERCOT grid. Using the synchrophasor data, these oscillations are evaluated with respect to location of the strongest signal (find the PMU closest to the source), the frequency and damping characteristics of the oscillation, and any identified interactions with other generation. The results are then tabulated, together with grid configuration metrics such as total amount of generation on-line, mix of generation, level of wind production, etc. In most of these cases, the source of the oscillations could be identified and confirmed through communications with the respective generation owners. In each case where a specific generator could be identified, the oscillations were confirmed to have been caused by incorrect settings in one of the generator control systems. This demonstrates the capability of the PMU monitoring system to be able to identify improperly performing control systems, and to enable the grid operators to contact the generation owners to effect corrective changes in the control settings. While this is not yet being accomplished in real time, the technology demonstrates the potential for being able to perform these analyses in near real time.

As additional PMUs are added to the ERCOT grid, the ability to detect, locate, and diagnose control-system caused oscillations will improve.

3.5.1.6 Event Analysis of Wind Interactions

Given that this demonstration project was focused on identifying interactions between wind generation and the ERCOT grid, and demonstrating improved capability to manage the grid with large amounts of wind generation operating, the first phase of the synchrophasor component efforts focused on expanding the PMU monitoring network and upgrading the computer network. At the same time, additional transmission had been under construction as part of the CREZ transmission expansion program to support the projected expansion in wind generation. The planned projects were expected to improve or add nearly 7,000 circuit miles of transmission lines and more than 17,000 megavolt amperes (MVA) of capacity to the grid. By the end of 2013, nearly all of the CREZ transmission expansion had been completed and placed into operation. This has facilitated construction of new wind generation across Texas.

At the beginning of 2010, there was 10,069 MW of wind generation installed on the ERCOT grid, and the peak wind production was recorded at 7,016 MW (June 2010). By the end of 2013, the ERCOT grid had more than 11,000 MW of installed wind capacity, the most of any state in the nation, and this was increased to 11,493 MW by October 2014. In November of 2014, a new ERCOT wind record was set at 10,301 MW which represented nearly 33% of the load at that time. Wind generation is forecast to increase to more than 18,500 MW installed.

Analysis of the recorded synchrophasor data for 2012 and 2013 was undertaken to examine potential interactions between wind generation and the ERCOT grid. Initial analysis identified low-level oscillations in the range of 5.0 – 5.5 Hz which are detectable at substations near the large wind generation areas in ERCOT. These oscillations are likely driven by control systems, as the frequencies are too high for typical electromechanical oscillations. It has also been observed that the oscillatory frequencies differ by region. More analysis was required to identify the grid conditions under which these oscillations appear, and whether they have the potential to interact with other generation. This study is described in Section 3.5.1.9.

Additionally, when unusual events occur on the grid, through the enhanced monitoring provided by the synchrophasor network, ERCOT is now able to quickly identify the nature of the event, and initiate corrective actions. An example of this enhanced monitoring and analysis capability occurred on January 10th, 2014, at about 8:30 a.m. After logging on to the RTDMS system, ERCOT Operations Engineers noticed oscillations on the RTDMS displays. Since the oscillations were showing up only at one location, a Wind Farm PMU, ERCOT looked at the generation at the Wind Farm unit close to the Wind Farm PMU. It was generating about 56 MW. These oscillations were showing up even though there was no line outage at the substation. (In October 2011, ERCOT had observed oscillations due to a line outage at the same Wind Farm substation).

In order to reduce the oscillations, ERCOT advised the wind farm unit to turn OFF their automatic voltage regulator. This did not help to reduce the oscillations. To reduce oscillations, the wind farm was curtailed to 45 MW output. This reduced the oscillations to some extent. Since the oscillations did not go away completely, the wind farm was further constrained to 40 MW and the oscillations finally went away. It was found that oscillations were present in the system since 6:15 p.m. of January 9, 2014 and lasted through the morning of January 10, 2014 until the unit was finally curtailed to less than 40 MW. Analysis of these oscillations using Phasor Grid Dynamics Analyzer (PGDA) tool indicated that the dominant mode that was present was 3.3 Hz. The voltage oscillations had a magnitude of about 1 kV. ERCOT Operations contacted the plant operator at the wind farm to determine the cause of oscillations. It was discovered that some updates were made to the settings for the system controller on January 9th, which matched the time of initial observation of the oscillations. After ERCOT informed the wind farm operators about the oscillations, the wind farm operator pulled back the updates which finally stopped the oscillations. After that, no oscillations were observed even when the plant was generating again at greater than 50 MW. A copy of the January 9-10, 2014 Wind Oscillation Event is attached as Attachment 8.

3.5.1.7 Event Analysis of Generator Control System Interactions

Another example of the benefits of the enhanced monitoring and analysis capability provided by the synchrophasor network occurred in mid-February, 2014, when ERCOT operations engineers noticed oscillations on the RTDMS displays when they first logged into the system in the morning. Upon analysis, it was found that the oscillations were showing up only on one PMU signal at the West 4 substation. ERCOT looked at the generation at the hydro unit close to West 4 PMU. It was generating about 25 MW. It was found that when the unit went offline at about 10:00 am, the oscillations died down. It was noticed that when the second unit at the same plant was running there were no oscillations observed. Oscillations showed up again on February 27th, 2014 when Unit 1 came online. Analysis of these oscillations using PGDA indicated that the dominant mode that was present was 1.8 Hz. ERCOT worked with the power plant operators to find the root cause of these oscillations with one of the units of the plant. It was confirmed that the plant operator could also see the oscillations for that unit. It was then decided that the plant operators would change some of the control cards for the problematic unit and test to see if it would solve the problem. Usually they operate the units alternately. Since ERCOT was going through some severe weather conditions, the plant operators decided to operate only the good unit until ERCOT passed through this severe weather conditions. On March 5th, when weather conditions were good, Unit 1 (for which some of the control cards had been replaced) was brought online at about 4:00 p.m. It was found that the oscillations were significantly reduced after the control cards were replaced. A copy of the Event Report is attached as Attachment 9.

3.5.1.8 Wind Characteristics - Inertial Frequency Response Study

Texas has the greatest amount of wind generation online in the nation and attains a new wind production record every year. Increasing wind production and penetration (percentage of total energy production) into the grid under different operating conditions poses operating challenges for ERCOT.

One of the challenges with high wind penetration is to maintain an adequate level of Inertial Frequency Response that is crucial to ensure reliable operation of grid. Inertial Frequency Response represents the inherent resistance of the grid to frequency decline following a sudden loss of generation. It is measured by the amount of generation loss (in MW) required to cause a first swing frequency decline of 0.1 Hz, and a larger value corresponds to increased grid resilience. This study was initiated to investigate the contribution of wind generation to the ERCOT grid Inertial Frequency Response based on a starting hypothesis that a decline in Inertial Frequency Response with respect to increasing levels of wind generation was being observed in the ERCOT Interconnection.

The study's analysis summarizes four illustrative scenarios that collectively concluded that Wind Generation provides no significant contribution to Inertial Frequency Response. Rather, Inertial Frequency Response appears to be primarily dependent upon the total amount of Non-Wind Generation available online, including non-wind generation that is unloaded but on-line to provide spinning reserve. The study report provides insights on the minimum amount of Non-Wind Generation needed to maintain adequate levels of Inertial Frequency Response under all conditions for the ERCOT Interconnection. It is attached as Attachment 10.

3.5.1.9 Oscillation Data Mining Study

As with every type of generation, both the characteristics of the generation and the associated control systems can cause local (and sometimes grid-wide) oscillations that have the potential to cause everything from power quality problems (e.g. flickering lights) to growing system-wide power swings and instability (e.g. WECC 1996 system breakup). Early detection, identification, and mitigation of oscillations prevent system vulnerability and customer complaints. Because wind generation is relatively new to the electric grid, its interactions with the grid are still being discovered. As the level of wind generation increases, oscillations which might be caused by improperly tuned control systems can possibly grow to high energy levels and interact with the rest of the grid. Baselining these grid oscillations is crucial to understanding their origins and ensuring reliable operation of grid.

As part of this effort, EPG has been retaining a full resolution record of all the synchrophasor data collected on the ERCOT grid since 2011. In order to better understand the types of oscillations that are present on the grid, and how those oscillations change over time, this Oscillation Data Mining study was undertaken to investigate oscillations in the ERCOT Interconnection and identify any interactions associated with increasing levels of wind generation.

This study focused on identifying unknown oscillations, determining the sources of these oscillations, determining the type of oscillation (wind related or driven by control systems), and baselining the oscillation characteristics (mode damping and energy). The study considered phasor data from January 2012 through mid-2014. PMUs near wind farms were selected for analysis. A Phasor Data Mining Application (PDMA) was created to perform the detailed scanning of the PMU data. The data scanning process identified "clean" data (using the C37.118 status information), then analyzed successive 60 second intervals of data for any oscillations with less than 8% damping. The results were then aggregated by frequency band, for each PMU, for each month. Analytics such as the level of occurrence during the month, and the highest oscillation energy recorded each month, were tabulated.

The study identified 10 distinct oscillatory modes in the ERCOT grid. Four of the modes were present throughout the two and a half year study period, and a fifth was present in both 2012 and 2013, but is no longer present in 2014. All 10 modes were compared to their corresponding recorded regional wind production to assess any relationship between the presence or energy level of the mode and wind generation.

The study concluded that one ongoing mode (0.9 Hz) appears to be related to the level of wind production, while the other three ongoing modes (5.0, 5.4, and 6.0 Hz) appear to be related to wind generator control systems. The full study is attached as Attachment 11.

3.5.1.10 Requirements for Control Room and Network Production Upgrades

During 2012 and early 2013, the minimum requirements for the computer and communications infrastructure needed to support the use of synchrophasor-based wide-area monitoring and visualization in the ERCOT control room were developed. Several possible configurations were examined, ranging from a single-instance implementation on production grade servers to a fully redundant implementation complete with staging and test environments. In early 2013, ERCOT decided to proceed with a single-instance implementation on production grade servers, with an associated network configuration. An illustration of the network configuration is shown below. EPG supplied an upgraded version of the RTDMS wide-area monitoring and visualization system, and had assisted ERCOT in configuring that version on the test system, in preparation for installation on the new production grade hardware. The new hardware was delivered in late July 2013 and the installation was completed in September 2013.

While ERCOT had been using the RTDMS monitoring capabilities within its Operations Engineering organization for several years to analyze grid events after-the-fact, in mid-2014 ERCOT migrated the RTDMS client software into the control room to provide the operators access to the real-time monitoring capabilities of RTDMS. ERCOT's operators are now able to detect emerging oscillations and perform post-event analysis after generating unit trips.

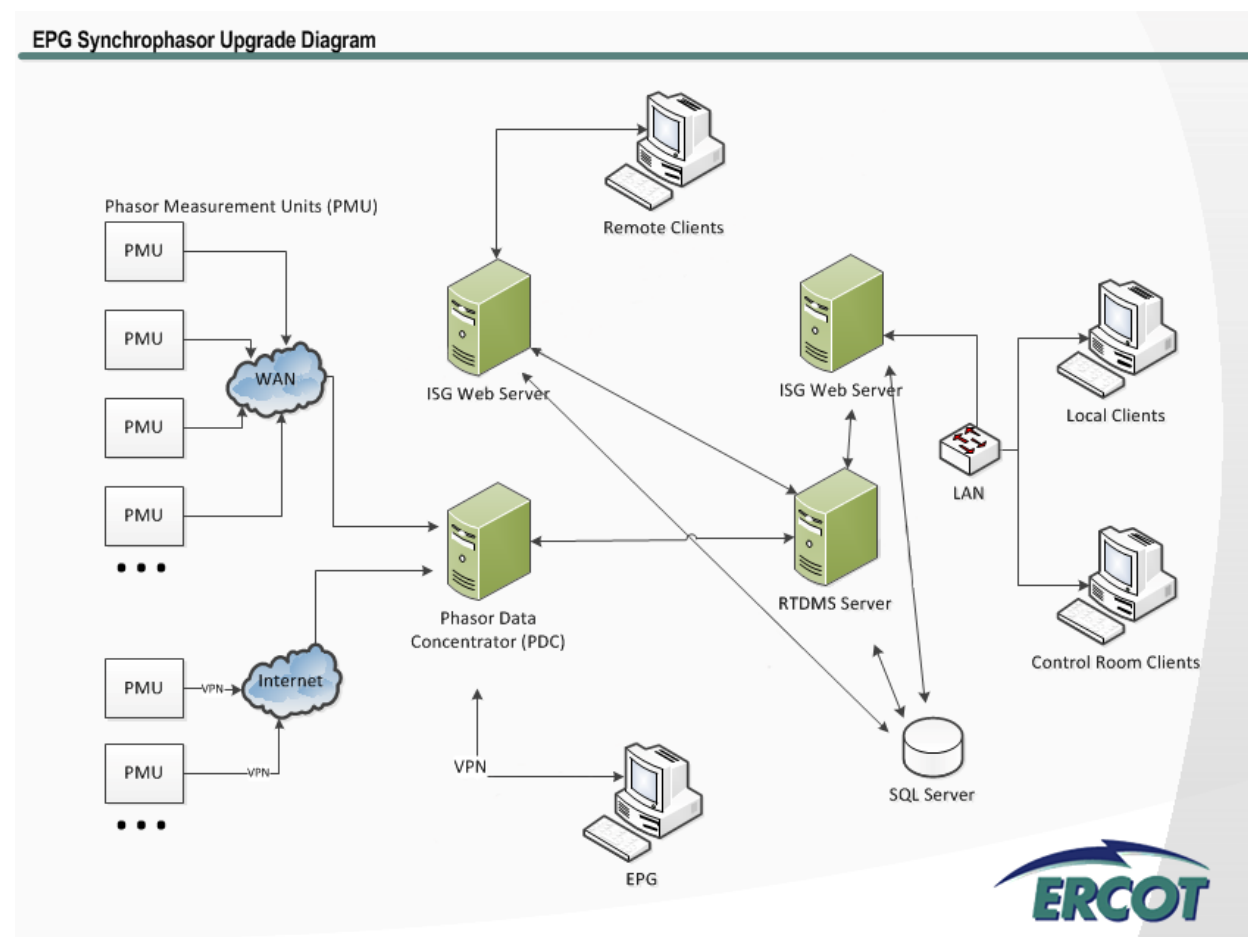


Figure 41. Synchrophasor Upgrade Diagram

3.5.1.11 Operator Training

In late September 2013, two on-site training sessions were conducted at ERCOT. A two-day training session covering the theory and practice of using the RTDMS wide area visualization system in real-time operations was provided to 38 operators from ERCOT and its member TDSPs. Additionally, a one-day training session covering the theory and practice of using the PGDA off-line analysis tool was provided to 23 operators and engineers from ERCOT and its member TDSPs.

In September 2014, a second two-day training session covering both RTDMS and PGDA was conducted at ERCOT. This training was provided to 44 participants, including both ERCOT operations personnel, and engineering and operations personnel from 17 ERCOT member organizations.

The training materials for both the 2013 and 2014 training sessions are posted on-line. The 2013 training materials are posted on the CCET Basecamp website. The 2014 training materials are posted on the EPG website. Participant review comments from both the 2014 RTDMS and PGDA training sessions are included as Attachment 12.

3.5.1.12 Generator Model Validation

The need for accurate generator dynamic modeling was highlighted in late 2011 during a scheduled transmission outage for maintenance. During the outage planning, ERCOT simulated the grid conditions that would be present during the outage and concluded that the wind generation connected to the portion of the grid where the outage would occur would remain stable; however, after removing the transmission line from service, severe voltage oscillations were experienced by the wind generator, resulting in loss of the generation. Utilizing the synchrophasor data recorded during the event, ERCOT planners were able to determine that the performance of the wind generation predicted by the dynamic outage planning studies did not match the actual performance of the wind generation. Changes to the dynamic model parameters were provided by the generation owner and incorporated into the ERCOT models. While ERCOT may be confident that this generator model is now accurate, there are hundreds of other generators that could be unintentionally misrepresented in ERCOT's dynamic studies. Utilizing synchrophasor data to validate predicted grid performance offers the potential of avoiding unexpected outages due to inaccurately modeled generation and load.

Synchrophasor data can be used to aid grid planners in validating the generator models upon which their dynamic simulation studies rely. Through separate CCET-sponsored research on behalf of ERCOT, researchers at the University of Texas at Arlington (UTA) developed algorithms that have the potential of directly validating the parameters that describe individual generators connected to the ERCOT grid by using synchrophasor data collected where the generator interconnects the grid. As part of this project, EPG, working in collaboration with UTA, implemented the generator parameter model validation algorithms into an off-line analysis tool. The algorithms have been tested with both simulated data and recorded synchrophasor data, to the extent possible, in order to demonstrate the potential for the algorithms to assist grid planners in validating their dynamic models. A summary of this new tool is attached at Attachment 13.

3.5.1.13 Slow Scan Synchrophasors

The Oncor EMS includes the capability to capture time-stamped voltage magnitude and voltage angle (synchrophasor) data in their substation Remote Terminal Units (RTUs), and to deliver that synchrophasor data into their Transmission Management System (TMS). Oncor has begun to implement this feature, bringing time-stamped data into their TMS at a rate of three samples per minute. This slow scan data is then delivered into the Oncor PI Historian using the Inter-Control Center Communications Protocol (ICCP). A Visual Studio Web service is used to process the data for display to the user. These displays include phase angles across transmission lines, phase angles between selected locations, phase angles relative to a selected location, and system stress using a relative phase angle dial indication. An illustration of the relative phase angle dial is shown below.

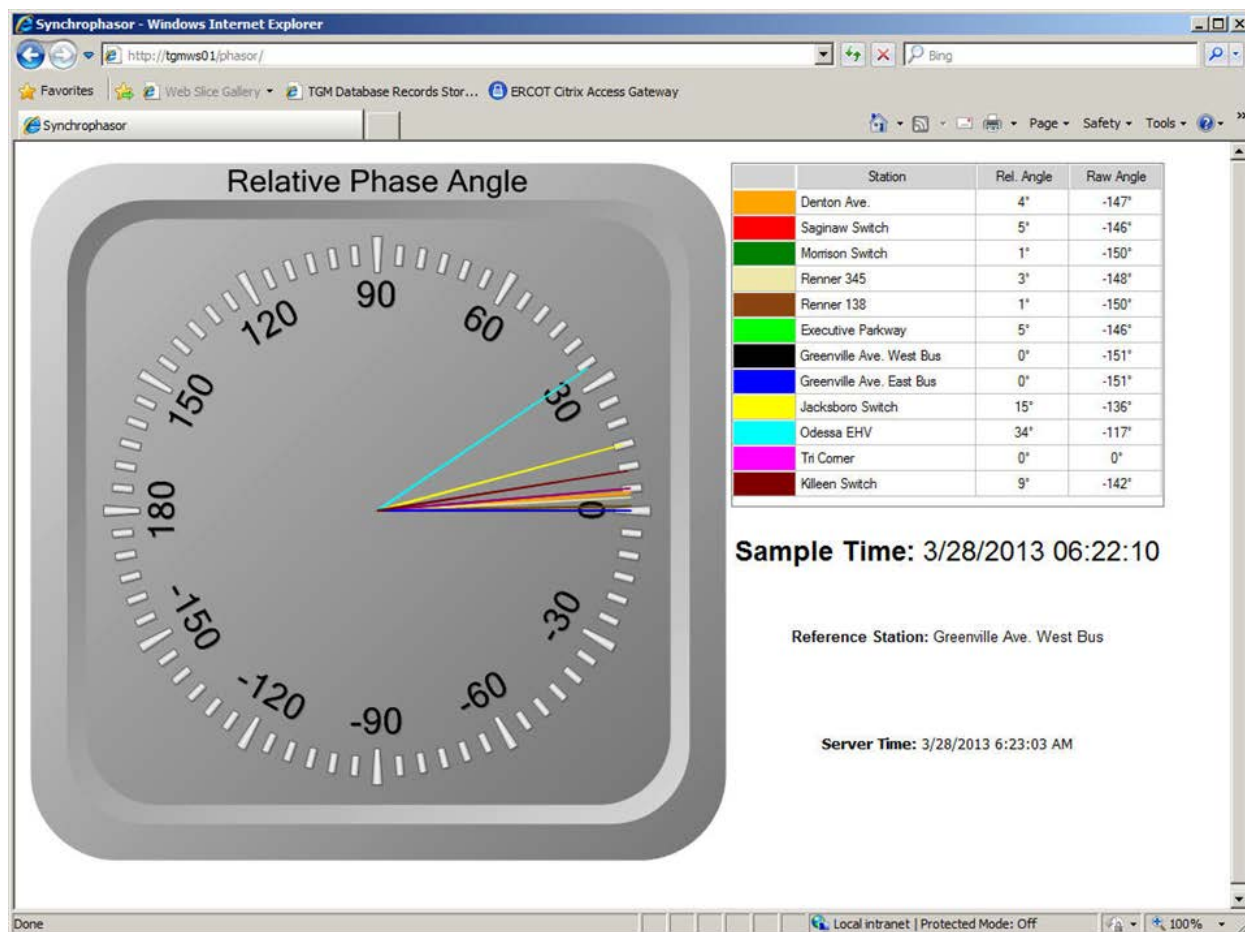


Figure 42. Relative Phase Angle Dial Illustration

Oncor is in the process of enabling this capability at more locations, stabilizing the data collection process, analyzing the data to establish alarm and notification limits, and developing a training program and operational procedures for deployment into the operational environment.

The key benefit of leveraging this slow scan synchrophasor data is the ability to monitor a significantly larger portion of the Oncor grid because the slow scan rate and limited data packet size make it possible to enable this capability at essentially every Oncor substation which is already using synchrophasor-enabled protective relays. The slow scan data cannot, however, be used to inform the operators about oscillatory events, nor can it be used to assist in validating dynamic generator models.

This slow scan data can potentially also be made available to ERCOT via the existing ICCP data link between Oncor and ERCOT. It has not yet been determined whether this expanded capability will be useful to ERCOT, in conjunction with the high speed synchrophasor data already being collected and displayed, or if this slow scan approach can be applied at other TDSPs.

3.5.1.14 Synchrophasor Use Cases

While the deployment of PMUs for the project was based on availability (PMUs were installed where the communications capability was sufficient, and the location was desirable for monitoring), part of the project plan included developing PMU location selection principles and criteria that can be used to site future PMUs. In support of this objective, the project team developed a set of 16 synchrophasor use cases as shown below. These use cases, summarized below, describe the grid operations application, grid scope, and the need. Over the past year or so, event analysis has been used to validate a number of these use cases. Of the sixteen identified use cases, ERCOT has already experienced and applied synchrophasor technology to nine of the use cases. Full details of these use cases and their supporting sample events are included as Attachment 14.

Table 13. Synchrophasor Use Cases

Use Case	Need	Grid Scope	Streaming 30 Samples/Sec	Slow Speed 3 Samples/Min	Local Event Capture
High Stress Across System (High Phase Angle) Observed	Wide area metric that uses phase angle as a measure of the degree of power flow from one region to another, or from one station to another. PMU phase angle data can advise the Shift Engineer about the measured angle across wide area to provide early warnings on high power flow (high grid stress)	Wide Area	Yes	Yes	
Small Signal Stability – Damping is Low	Small signal oscillation is generally across a large area of the grid, and can be noted with a low damping ratio. PMU data can advise the Shift Engineer about both known and unknown oscillations at location(s)	Wide Area	Yes		
Small Signal Stability – Emerging Oscillation Observed	Small signal oscillation is generally across a large area of the grid, and can be noted as a new oscillation frequency or mode not seen before. PMU data can advise the Shift Engineer about both known and unknown oscillations at location(s)	Wide Area	Yes		
Voltage Oscillation Observed	This is a regional phenomenon in one area of the ERCOT grid. The cause may be over-control of generator exciters or reactive controls in low electrical strength areas of the grid. PMU voltage phasor can advise the Shift Engineer about the voltage oscillations at location(s) due to fast voltage controllers at wind generators and other control devices in the grid	Regional	Yes		

Voltage Instability Monitoring (real-time P-V or Q-V curve)	This involves regional voltage instability based on deteriorating grid conditions noted with real-time tracking of P-V or Q-V curves. PMU data (real, reactive power and voltage) can advise the Shift Engineer indirectly on high grid stress under low voltage deteriorating conditions	Regional	Yes		
Detection of Subsynchronous Interactions (not necessarily resonance, just below 60 Hz)	Detection of subsynchronous (<60 Hz) interactions at a higher frequency than those with small signal stability or emerging oscillation. Oscillation frequency can advise Shift Engineer	Local Regional	Yes		
Integrate PMU Data into State Estimator	PMU phase angles can be used to validate the state estimator results used in control rooms (locates differences which reflects anomalies in models used for state estimation)	Wide Area	Yes	Yes	
System Disturbance – Capture and Interpretation	PMU data is useful for event analysis and for determining the root cause of events and their locations	Regional	Yes	Yes, not high resolution	Yes
Generator Parameter Determination	Generator parameters in actual settings versus models used in stability studies don't always match actual transient behavior in system events. PMU data (voltage, phasor, P&Q) can advise generator dynamic response following a nearby transient, compares results to simulated response (based on system planning models), and alerts if differences are significant (meaning that the generator response to the transient event was different from what was expected)	Local	Yes		Yes
Major Load Parameter Determination	Similar to generator parameter determination, but this addresses load models that behave differently than models being used.	Local	Yes		Yes

PMU-Based Fault Location	PMU data is used to detect a fault on the network, and algorithms provide grid operator with estimate of likely location of the fault, and the fault characteristics. Shift Engineer reviews fault information and determines need for reclosing test	Local Regional	Yes		Yes
Phase Angle Across Breaker for Reclosing Action	PMU data is useful during an event to identify stress across system, and validate safe restoration actions. System operator reviews phase angle and voltage on both sides of an open breaker and determines need and ability to successfully reclose the breaker		Yes	Yes	
Synchronous Resonance Identification and Mitigation	ERCOT planning guide revision request (PGRR) 027 needs subsynchronous resonance (SSR) detection algorithms related to series capacitors to determine what capacitors to swap out	Regional	Yes		
Transmission Characteristics Determination	PMU voltage magnitude and phase angle measurements at the two ends of a transmission line can be used to compute the characteristics of the transmission line, and to validate the modeling used for studies	Regional	Yes		Yes
Dynamic Transmission Line Ratings using PMU Monitoring	PMU data will be used to monitor the change in transmission line resistance versus temperature. For example, at x temperature, the resistance deviation is y at this z sag level. The temperature will be independent and used in an offline study which will use apparent resistance to determine the sag	Regional	Yes		
Validation of Control Devices (e.g. SVC) Performance	PMU data can be used to validate the performance of control devices, such as SVCs, following grid events. The Shift Engineer reviews the measured performance, and compares it to expected performance	Regional	Yes		Yes

3.5.2 Metrics and Benefits Analysis

The anticipated benefits of development and deployment of a synchrophasor network in the ERCOT grid are focused around improving the grid operators' ability to manage large amounts of wind generation.

3.5.2.1 Operational Benefits

There is a wide array of potential operational benefits through the deployment of PMUs, some of which include:

- Identification of emerging grid conditions which, if not properly mitigated, lead to loss of generation or load
- Identification and correction of undesirable generation interactions, such as oscillations
- Identification of discrepancies between measured and predicted grid response to events, including frequency bias and dynamic response, generator dynamic response, and load dynamic response.

Benefits identified to date (and described previously) have included detection and identification of undesirable performance of generator control systems, and generator dynamic modeling discrepancies (differences between predicted and actual performance). While loss of generation occurred in at least two of these identified events, the generation lost was the source of the discrepancy, and no customer load was lost.

In one incident in early 2014, ERCOT detected that a specific generating unit was causing low level oscillations on the grid, which was adversely impacting customer power quality. At the request of ERCOT, the generator operator agreed to avoid use of the misbehaving generator during a high load period, so as to avoid adverse power quality impacts. Subsequently, upon replacement of a control component within the generator, ERCOT was able to use its real-time synchrophasor monitoring system to verify that the oscillations had been eliminated, allowing release of the operating limits on the offending generator.

In each of these events, no customer load has been lost, thus, quantification of benefits, such as mitigation costs or economic loss, is a challenge at this time.

3.5.2.2 Build Metrics and Benefits

The average cost of purchasing, installing and supporting synchrophasor equipment is difficult to quantify since these costs are impacted by many variables, not the least of which are:

- Is this a new PMU or is it an upgrade to an existing PMU?
- Is this the first PMU at the chosen location or are others already in place?
- Are appropriate high-speed communications already available, or will they be required?
- Are secure communications paths available, or will they be required?

- Does the TDSP already have the requisite servers to support the monitoring software and data archiving needs, or will they be required?
- Are the costs (being donated as cost share contributions) limited to certain aspects of an installation?
- How much did the costs increase from 2010 to 2014 (long-term project)?

Considering that each installation is different, the following provides some basic cost information related specifically to this project.

- The average cost of a PDC was \$13,374, but these costs ranged from \$7,500 (software only) to \$28,350 (hardware, software, IT support).
- The average cost of a PMU was \$23,835, but these costs ranged from \$13,020 (multiple hardware units only) to \$247,278 (including communications and field installation).
- A typical ePDC license cost ranged from \$16,000 to \$30,000 but this was vendor specific.
- Monitoring software costs can be priced by seat, by number of users, or as part of an annual maintenance and support contract, and different customers may receive discounts. These costs ranged from \$10,000 (discounted in 2010) to \$32,000 (in 2014).

3.5.2.3 Impact Metrics and Benefits

The benefits derived from the installation of a synchrophasor monitoring system are primarily associated with improved real time grid monitoring, improved grid modeling capability, improved understanding of the performance of the grid (e.g. baselining and post event analysis), and improved diagnostic capabilities. While each of these has the potential to reduce overall costs, the actual savings are difficult to assess.

- Improved real time grid monitoring – as illustrated in Section 3.5.1.6, using the synchrophasor system, the ERCOT grid operators were able to detect and proactively manage an emerging oscillation condition on the grid, preventing potential loss of generation and/or unacceptable customer power quality degradation
- Improved grid modeling capability – as noted in Section 3.5.1.12, the ability to use synchrophasor data following routine grid events to verify and, if needed, adjust the parameters used to model generating units, enables the grid operator to improve the accuracy of the modeling tools used to establish all the operational limits on the grid. With more accurate operational limits, power delivery (efficiency) may be optimized while reliability is enhanced.
- Improved understanding of the performance of the grid – through the use of baselining analysis, the monitoring and alarming capability of the synchrophasor network can be tuned to alert the grid operator to abnormal operating conditions, providing both an early warning system and backup monitoring capability to the energy management system. For instance, the baselining studies have confirmed that in the event of an outage of the energy management system, the synchrophasor voltage angle spreads between substations provides an accurate indication of the power flows on the network. Moreover, the synchrophasor network provides enhanced ability to analyze and evaluate the grid performance following disturbances. These two capabilities enhance the overall reliability of the grid, and reduce the potential for unneeded

curtailments and outages during abnormal operating conditions, thus improving the efficiency of the grid operation.

- Improved diagnostic capabilities – the synchrophasor system enables the automatic capture of high resolution data, which enhances the grid operator’s ability to identify all the precursor events leading up to a grid event such as loss of generation or transmission. This high resolution data enables the operators to quickly and easily sequence the events (while the EMS can identify all the switching events which occur, the EMS timestamp is generally not sufficiently accurate to provide sequencing information), providing a clear event sequence timeline. Often, this higher resolution data is able to identify precursor events which might otherwise be more difficult to link to the event. This enhanced diagnostic capability both reduces the time needed to perform detailed analyses, and improves the accuracy.

3.5.3 Stakeholder Feedback

The TDSPs have indicated significant benefits being provided by synchrophasors, and are continuing to expand the number of PMUs in their territories. In late 2013, ERCOT formed a phasor measurement task force, with the intent of developing protocols and guides for incorporating the synchrophasor monitoring system into its control room operations. As noted earlier, this task force has recently completed its work, and its protocols and guides (including the Communications Handbook) are now available for ERCOT member systems to utilize. Additionally, ERCOT has provided remote access to its grid monitoring displays so that the TDSPs can visualize what’s happening not only in their territory but across the ERCOT grid.

3.6 Conclusions

This section covers projections of demonstration and commercial scale system performance as well as lessons learned and best practices.

3.6.1 Projections of Demonstration and Commercial Scale System Performance

With the experiences gained in measuring and monitoring the ERCOT grid using synchrophasors, ERCOT has moved this technology into their control center, and is using the results not to define actions, but to help inform them about the performance of the grid, and emerging conditions on the grid. The benchmarking and baselining activities have provided the confidence that the synchrophasor-based monitoring accurately represents the condition of the ERCOT grid, and thus can be relied upon during brief outages of the EMS system. Moreover, the synchrophasor monitoring system is successfully identifying emerging conditions on the grid, such as control-system caused oscillations. This has enabled ERCOT operators to quickly identify the source of the oscillations and mitigate their impacts. As the monitoring system continues to expand coverage of the ERCOT grid, it is expected that the synchrophasor-based measurement of grid performance following disturbances will be increasingly utilized to enable validation and improvement of engineering modeling of the grid.

3.6.2 Lessons Learned and Best Practices

There are several lessons learned as part of development of the ERCOT synchrophasor network. First and foremost is to monitor, measure and validate that the synchrophasor data being sent from the substation is faithfully and reliably delivered through the communications network and into the central data base (such as at ERCOT). During the early development of the ERCOT synchrophasor network, the team was focused on connecting additional PMU units, and initiating the data flows. Validating that the data was being reliably delivered all the way to the RTDMS data base (at ERCOT) was a secondary goal. Moreover, without reliable monitoring tools, neither the TDSPs nor ERCOT had easy visibility of the day-to-day synchrophasor data flow across the network. The first step in ensuring that the data was flowing reliably was to modify the visualization software to provide an automated means for preparing daily monitoring reports and then distribute those reports to the TDSPs and ERCOT. Next was to validate that the monitoring report was accurate. Lastly, the data being archived (which was being used to create the monitoring report) was validated against the data being sent. Completion of this enabled the team to confidently rely on the daily performance reports.

The second major lesson learned was the importance of benchmarking the results of the synchrophasor data with the EMS data, to validate that both systems produced comparable results. This was accomplished initially as part of the 2012 Baselining Study.

A third lesson learned was to perform clustering analyses, to validate that PMUs which appeared to be geographically (and electrically) near each other did, in fact, perform similarly. This also enabled ERCOT to select representative PMUs from within a cluster for real-time monitoring purposes.

Finally, through the performance of baselining studies, alarm limits could be established for use in the RTDMS synchrophasor monitoring system which would effectively alert the grid operators when conditions were nearing, or exceeding limits based on the synchrophasor measurements, independently from the EMS system alarms. These baselining studies also confirmed that the grid stresses, and corresponding alarm limits, did indeed change when new transmission lines were placed in service (as was the case during 2012 and 2013).

Replication of these lessons in other grids is expected to improve the performance of the synchrophasor network, enhance the confidence in the resulting measurements, and improve the overall cost/benefit relationship.

4. SYNCHROPHASOR SECURITY FABRIC

4.1 Introduction

Another synchrophasor demonstration effort focused on incorporating enhanced security protection mechanisms in the synchrophasor network. For this effort, Intel/McAfee provided Security Fabric (SF) components that are architected to address the seven security tenets of the NIST-IR 7628 for synchrophasor networks. EPG, a member of the CCET Team, integrated those capabilities with its monitoring, visualization and reporting platforms (ePDC, RTDMS, and Security Fabric Gateway) to demonstrate secure data transport and data visualization.

One of the key concepts of the Security Fabric approach is the separation of protection and security. Protection is a mechanism while security is a policy. Therefore, making the protection-security distinction for separation of mechanism and policy principle is important. Protection is provided as a fault tolerance mechanism by hardware/firmware and kernel whereas the operating system and applications implement their security policies. In this design, security policies rely therefore on the protection mechanisms and on additional cryptography techniques.

The technical objectives of this effort included:

- Intrusion detection using protocol whitelisting
- Attack detection and remediation
- Secure encryption communications operating efficiently at all levels
- Audit logging and real-time correlation
- Upgrades for devices already operating in the field

The project employed contemporary concepts of silicon-driven security to stop attacks that easily defeat software-only security attempts. This provides a more secure framework for incorporating security in the physical hardware layers to better protect sensitive operations i.e. providing authentication for every step of the boot sequence to eliminate infection by malware or root-level viruses. The project approach leveraged the hardware security, and provided whitelist attestation to verify the trust basis for all changes and transition control to promote changes at a safe and convenient time.

For this demonstration, EPG designed, integrated, validated, and tested the Security Fabric enhancements to its ePDC and RTDMS applications in its laboratory environment. This testing was completed in early December 2013. In mid-December 2013, the Security Fabric servers together with the EPG software were deployed at TTU. A field demonstration was performed on the TTU synchrophasor network to demonstrate the use of the Security Fabric to secure the phasor data stream between a “virtual Transmission Owner/Utility” and a “virtual Independent System Operator”. The goal of the demonstration was to validate that the phasor data stream could be successfully secured from

cyber attack while still delivering timely phasor data to the “virtual ISO’s” wide-area monitoring and visualization system. As a part of the demonstration, TTU faculty and students performed compliance and penetration testing of the Security Fabric system, and documented their findings. Their assessment is included as Attachment 16.

4.2 Anticipated Benefits

Broadly, the expected benefits of the CCET demonstration project are to be derived from a more reliable electric grid that can facilitate effective management of and responses to increased wind resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage. Customers are to be empowered to effectively and reliably manage their peak demand, therefore resulting in reduced customer electricity costs, reduced system-wide capacity needs, reduced electrical losses, and reduced environmental impacts. Specific smart grid benefits that are potentially supported by the synchrophasor Security Fabric component include:

- Application whitelisting on management systems
- Movement of databases to secured management system with encrypted tunnel
- C37.118 data (standard for PMU accuracy and data communications) is only received from authorized nodes
- C37.118 data is exchanged between nodes using a secure tunnel
- Access to applications is limited to authorized clients
- Secure network time protocol communications

4.3 Interactions with Key Stakeholders

The Security Fabric suite of capabilities is still evolving, but its first set of capabilities was intended to provide a requisite level of cyber security protections without impacting existing operations or performance, and without having to modify the applications requiring protection. While development and testing was ongoing, interaction with stakeholders was limited, but once the testing was completed, a number of stakeholders expressed keen interest in the potential for not only securing synchrophasor networks but any network requiring cyber security protection schemes.

4.4 Technical Approach

This section provides a detailed description of the technical approach, the questions to be addressed, and the methods used to answer them. The first part, Project Plan, includes illustrations of the system configurations, defines initial steps to achieve effective deployment, identifies the steps taken to demonstrate the technology (including test procedures), and outlines the approach to conduct analysis to assess performance. The second part, Data Collection and Benefits Analysis, defines the technical approach for conducting the benefits analysis, including the methodology and algorithms when appropriate.

4.4.1 Project Plan

The Security Fabric project was intended to demonstrate the capability to securely deliver synchrophasor data between two separate entities (e.g. a TDSP and ISO) while defending against cyber-attack. This demonstration was developed in three phases.

In the first phase, several PMUs installed near TTU (on the campus, at the RTC, and in nearby utility substations), delivered their synchrophasor data to an ePDC located at the RTC. One of the unique features of this network is that it directly monitors the output of several wind generators connected to the local distribution utility network. This enables “close-up” monitoring of any wind-related interactions with the grid. The ePDC at the RTC represented, for the purposes of this demonstration project, a “virtual TDSP” PDC. The time-aligned synchrophasor data from the RTC ePDC was then streamed over the TTU communications network to an ePDC located in the Engineering Building on the TTU campus, and from there into the RTDMS wide-area monitoring and visualization system. The equipment located in the Engineering Building represented, for the purposes of this demonstration project, a “virtual ISO”. At TTU, the synchrophasor data was displayed in RTDMS and stored in a phasor archiver for post-event analysis using the PGDA. This first phase has been operational since early 2013.

In the second phase, a test configuration consisting of a Security Fabric-enabled ePDC (SF-ePDC) together with a Security Fabric-enabled RTDMS (SF-RTDMS) was established in the EPG laboratory in Pasadena, CA. A simulated data stream was delivered into the SF-ePDC, and the SF-RTDMS was configured to provide wide-area monitoring and visualization of that data. This test system was then subjected to cyber-attack to validate the design and performance of the Security Fabric elements, monitor the impact of the Security Fabric on the overall performance of the synchrophasor tools (SF-ePDC and SF-RTDMS), and measure the latency introduced by the Security Fabric. The second phase was completed in late 2013 with a successful demonstration, in the EPG laboratory, of the Security Fabric-enabled ePDC and RTDMS applications maintaining normal operation while withstanding simulated cyber-attacks.

In the third phase, a second instance of the ePDC, equipped with the SF-ePDC, was installed at the RTC, and a synchrophasor data stream from the first ePDC was delivered to the SF-ePDC. The output synchrophasor data stream from the SF-ePDC was directed to the SF-RTDMS located in the Electrical Engineering Building on the TTU campus. Like the first instance of RTDMS, this SF-RTDMS also provides wide-area monitoring and visualization, and the data stream is stored in a phasor archiver. Startup testing of this parallel configuration, shown below, was completed in December 2013, and TTU collaborators spent much of 2014 performing compliance and penetration testing on the SF-enabled network, and side-by-side comparative performance testing, the results of which are captured in this report.

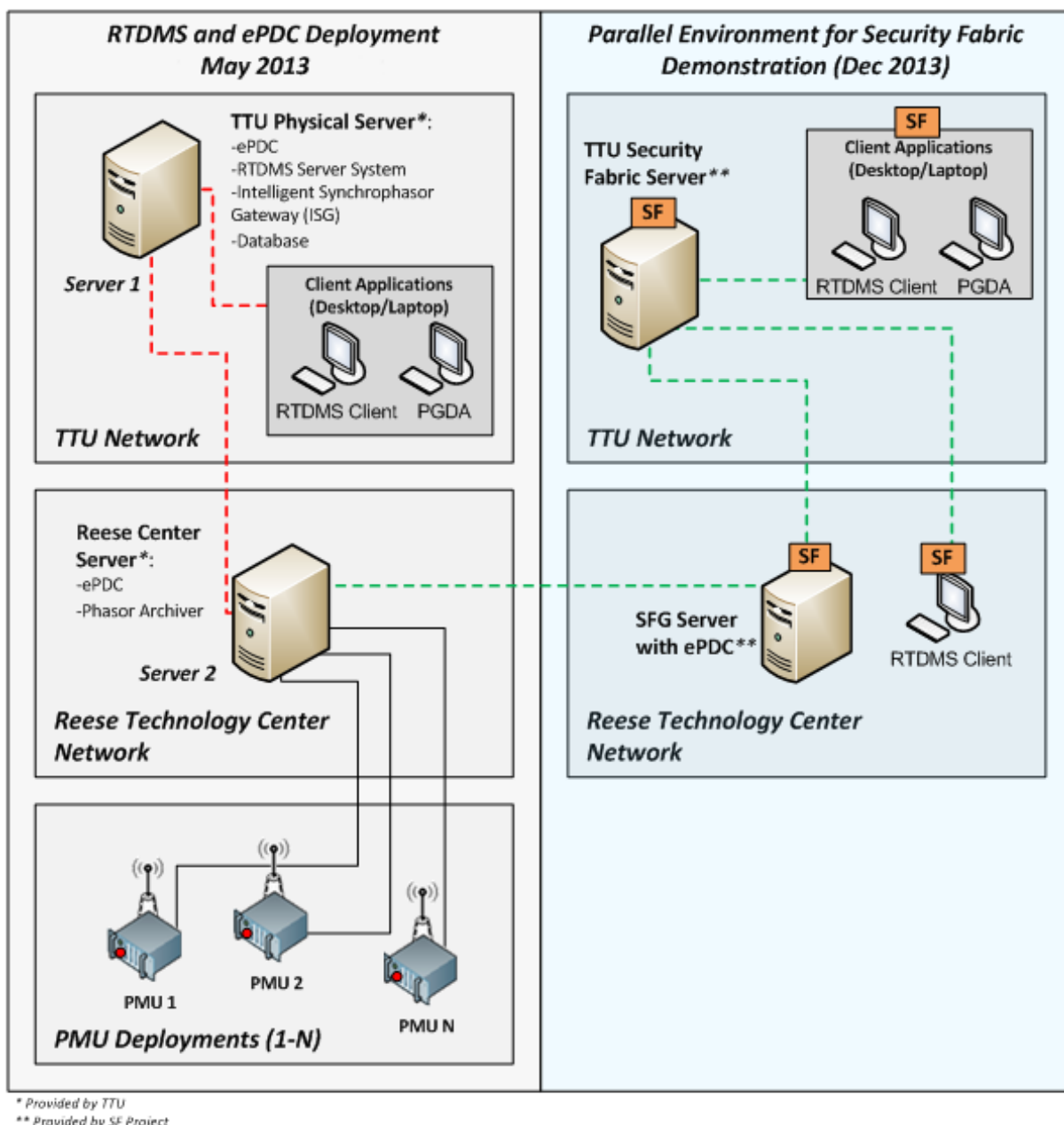


Figure 43. Basic Synchrophasor Network and the Security Fabric-enabled Network at TTU

While intentional cyber-attack testing initiated by TTU collaborators was planned, the Security Fabric has already proven its voracity by monitoring, defending against, and logging numerous attempts to gain access to various components in the overall system from the external Internet. The initial results and analysis of the Security Fabric monitoring of these intrusion attempts is included in Attachment 15.

As part of the third phase, the security testing performed at TTU consisted of two parts, both of which were black-box testing.

- The first part was ***verification of security fabric requirement specifications***. The study of security requirements, acquisition of appropriate requirements and development of security specifications are critical to thorough testing results. This effort involved testing to determine whether the Security Fabric-enabled network at TTU satisfied the security requirement specifications provided by Intel/McAfee. The specifications were based on the “Guidelines for Smart Grid Cyber Security” NISTIR 7628 requirements (http://www.nist.gov/smartgrid/upload/nistir-7628_total.pdf). The verification of the specifications iterated through evolving changes of the requirement specifications.
- The second part was ***penetration testing*** on both the synchrophasor network without Security Fabric and the Security Fabric-enabled Network at TTU. This involved coding scripts to identify security vulnerabilities and injecting attacks by exploiting selected vulnerabilities found. The details of specific types of tests performed for both parts along with the results are described in Section 4.5.1 and Attachment 16.

4.4.2 Data Collection and Benefits Analysis

To support the validation and analysis portions of the Security Fabric component, the following data was collected, retained, and analyzed:

- Synchrophasor data was retained in full resolution for the entire demonstration period from both the phase one ePDC-RTDMS network and the phase three SF-ePDC – SF-RTDMS networks. This data was used to monitor the overall performance of the TTU PMU network, including the communications network and the computer network infrastructure. Data from the two phases was compared to validate that the data flowing through the SF-enabled network was unaltered by the Security Fabric components.
- Grid events, such as loss of generation or load, severe voltage dips, or sustained oscillations that were detected by the TTU RTDMS wide-area monitoring and visualization system were also retained and analyzed.

Additionally, the full synchrophasor data was scanned for unusual signatures, such as low-level oscillations which are characteristic of electromechanical interactions between generation and load, as well as control-system driven oscillations. Because the TTU PMU network was monitoring a distribution network with both local thermal and wind generation, it was expected that interactions between different types of generation might be more visible in this network.

Performance metrics from both the phase one network and the SF-enabled network were compared both during normal operation and during cyber-attack testing to identify and calibrate the impact of the Security Fabric on the overall delivery of synchrophasor data between the SF-ePDC and the SF-RTDMS components.

4.5 Performance Results

This section covers the operation of smart grid technologies and systems, including the demonstration results; impact metrics and benefits analyses; and any stakeholder feedback received to date.

4.5.1 Operation of Smart Grid Technologies and Systems

The functional goal of the Security Fabric demonstration was to demonstrate delivery of a synchronized PMU data stream over a secure communications framework for the Internet Protocol (IP) interconnection between two locations: a Security Fabric-enabled ePDC at the Reese Technology Center connected to the Independent System Operator's ePDC (represented by an ePDC located at TTU). The functional goal of this demonstration for data visualization security was to enable an RTDMS client and RTDMS server with Security Fabric to provide secure data visualization for the TTU synchrophasor system.

The first phase of this demonstration was the establishment of a synchrophasor network at Reese Technology Center and TTU. For this first phase, the ePDCs at Reese and TTU were both in place and operational. The RTDMS is operational at TTU, along with an RTDMS Client, which provides the wide-area monitoring and visualization capabilities. Phasor Archivers (to store the synchrophasor data stream) are installed and operational. Four PMUs are currently delivering data to the Reese Technology Center. This represented the baseline system against which the performance of the Security Fabric-enabled system was compared when the additional Security Fabric-enabled components were installed at TTU.

The results of security testing performed at TTU are described in the next two subsections.

4.5.1.1 Summary of Security Fabric NISTIR Requirement Specifications Testing Results

Intel/McAfee provided Security Fabric relevant security specifications that cover 11 categories of the NISTIR 7628 security requirements. The Security Fabric-enabled network at TTU was tested against these specifications. The process of testing in this part involved three basic activities: (1) *requirement acquisition* (communicating and assisting in the interpretation of the testing requirements according to NIST standards), (2) *specification description* (providing feedback and clarification in defining proper security specifications), and (3) *specification verification* (verifying that the system satisfies the given security specifications). This process is time consuming due to the nature of evolving requirements and specifications development.

The table below shows the respective specifications by category and a general indication of whether or not that requirement was satisfied through testing for each of the 74 specifications. The table results indicate that 86.5% of these security specifications were satisfied. By being diligent, and testing beyond the specifications, the TTU Team determined that the remaining requirements could be satisfied by 1) clarifying the specification so that the test intent, objectives, steps and results are adequately defined,

and 2) incorporating additional existing Intel/McAfee tools or applications. The detailed specifications, their corresponding verification results and suggested revisions are provided in Appendix A of Attachment 16.

Table 14. Overview of Verification Result

Spec. #	Requirement	Result
Category 1: Access Control (SG.AC)		
1	SG.AC-2 Remote Access Policy and Procedures	Satisfactory
2	SG.AC-3 Account Management	Need Clarification
3	SG.AC-4 Access Enforcement	Satisfactory
4	SG.AC-5 Information Flow Enforcement	Satisfactory
5	SG.AC-6 Separation of Duties	Need Clarification
6	SG.AC-7 Least Privilege	Need Clarification
7	SG.AC-8 Unsuccessful Login Attempts	Satisfactory
8	SG.AC-9 Smart Grid Information System Use Notification	Satisfactory
9	SG.AC-10 Previous Logon Notification	Need Clarification
10	SG.AC-12 Session Lock	Satisfactory
11	SG.AC-13 Remote Session Termination	Satisfactory
12	SG.AC-15 Remote Access	Satisfactory
13	SG.AC-18 Use of External Information Control Systems	Satisfactory
14	SG.AC-19 Control System Access Restrictions	Satisfactory
15	SG.AC-20 Password	Satisfactory
Category 2: Audit and Accountability (SG.AU)		
16	SG.AU-2 Auditable Events	Satisfactory
17	SG.AU-3 Content of Audit Records	Satisfactory
18	SG.AU-4 Audit Storage Capacity	Satisfactory
19	SG.AU-5 Response to Audit Processing Failure	Satisfactory
20	SG.AU-6 Audit Monitoring, Analysis, and Reporting	Satisfactory
21	SG.AU-7 Audit Reduction, and Report Generation	Satisfactory
22	SG.AU-8 Time Stamps	Satisfactory
23	SG.AU-9 Protection of Audit Information	Satisfactory
24	SG.AU-10 Audit Record Retention	Satisfactory
25	SG.AU-11 Conduct and Frequency of Audits	Satisfactory
26	SG.AU-14 Security Policy Compliance	Satisfactory
27	SG.AU-15 Audit Generation	Satisfactory
28	SG.AU-16 Non-Repudiation	Satisfactory
Category 3: Security Assessment and Authorization (SG.CA)		
29	SG.CA-4 Smart Grid Information System Connections	Satisfactory
30	SG.CA-6 Continuous Monitoring	Satisfactory
Category 4: Configuration and Management (SG.CM)		
31	SG.CM-6 Configuration Setting	Satisfactory
32	SG.CM-7 Configuration for Least Functionality	Satisfactory
33	SG.CM-8 Component Inventory	Satisfactory
Category 5: Continuity of Operation (SG.CP)		
34	SG.CP-10 Recovery and Reconstitution	Satisfactory

Spec. #	Requirement	Result
35	SG.CP-11 Fail-Safe Response	Satisfactory
Category 6: Identification and Authentication (SG.IA)		
36	SG.IA-3 Authenticator Management	Satisfactory
37	SG.IA-4 User Identification and Authentication	Satisfactory
38	SG.IA-5 Device Identification and Authentication	Satisfactory
39	SG.IA-6 Authenticator Feedback	Satisfactory
Category 7: Incident Response (SG.IR)		
40	SG.IR-6 Incident Monitoring	Satisfactory
Category 8: Development and Maintenance (SG.MA)		
41	SG.MA-2 Legacy Smart Grid Information System Upgrades	Satisfactory
42	SG.MA-6 Remote Maintenance	Satisfactory
Category 9: Risk Assessment (SG.RA)		
43	SG.RA-4 Risk Assessment	Satisfactory
44	SG.RA-6 Vulnerability Assessment and Awareness	Need Additional To
Category 10: Communication Protection (SG.SC)		
45	SG.SC-2 Communications Partitioning	Satisfactory
46	SG.SC-3 Security Function Isolation	Satisfactory
47	SG.SC-4 Information Remnants	Satisfactory
48	SG.SC-5 Denial-of-Service Protection	Satisfactory
49	SG.SC-6 Resource Priority	Satisfactory
50	SG.SC-7 Boundary Protection	Satisfactory
51	SG.SC-8 Communication Integrity	Satisfactory
52	SG.SC-9 Communication Confidentiality	Satisfactory
53	SG.SC-10 Trusted Path	Satisfactory
54	SG.SC-11 Cryptographic Key Establishment and Management	Satisfactory
55	SG.SC-12 Use of Validated Cryptography	Satisfactory
56	SG.SC-15 Public Key Infrastructure Certificates	Satisfactory
57	SG.SC-18 System Connections	Satisfactory
58	SG.SC-19 Security Roles	Satisfactory
60	SG.SC-20 Message Authenticity	Satisfactory
61	SG.SC-21 Secure Name/Address Resolution Service	Not Applicable
62	SG.SC-22 Fail in Known State	Need Clarification
63	SG.SC-23 Thin Nodes	Need Clarification
64	SG.SC-24 Honeypots	Not Applicable
65	SG.SC-25 Operating System Independent Application	Satisfactory
66	SG.SC-26 Confidentiality of Information at Rest	Satisfactory
67	SG.SC-27 Heterogeneity	Satisfactory
68	SG.SC-28 Virtualization Technique	Satisfactory
69	SG.SC-29 Application Partitioning	Satisfactory
70	SG.SC-30 Smart Grid Information Partitioning	Satisfactory
Category 11: Information Integrity (SG.SI)		
71	SG.SI-2 Flaw Remediation	Need Additional To
72	SG.SI-4 Monitoring Tools and Techniques	Satisfactory
73	SG.SI-6 Security Function Verification	Satisfactory
74	SG.SI-9 Error Handling	Satisfactory

4.5.1.2 Summary of Penetration Testing Results

The figure below shows overview results of the penetration testing performed at TTU on the synchrophasor network *without Security Fabric*. The right of the figure summarizes different types of vulnerabilities found along with their corresponding possible attacks/effects. The first two vulnerabilities could result in man-in-middle attacks (hacker makes independent connections with, and relays data to, each end point of a process). On the other hand, the database open access vulnerability came from the violation of PCI DSS (https://www.pcisecuritystandards.org/security_standards) whose consequence of successful exploitation would allow an attacker to steal information from the database. As shown in the table, the vulnerabilities are categorized by their causes: *misconfiguration of protocol* and *misconfiguration of policy*. The diagram on the left of the figure shows where the vulnerabilities were found. Each circle represents a specific vulnerability found for each specific service. For example, the *SMB signing disabled* vulnerability was found in two of the ePDC services and in one of the RTDMS service components contributing to three vulnerabilities out of a total of six vulnerabilities in the TTU synchrophasor system.

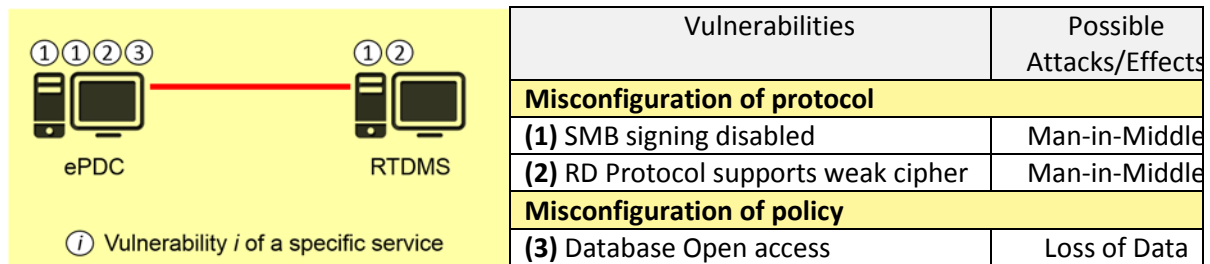


Figure 44. Vulnerabilities in the TTU Synchrophasor System Without SF

The next figure shows the vulnerabilities detected in the Security Fabric-enabled network.

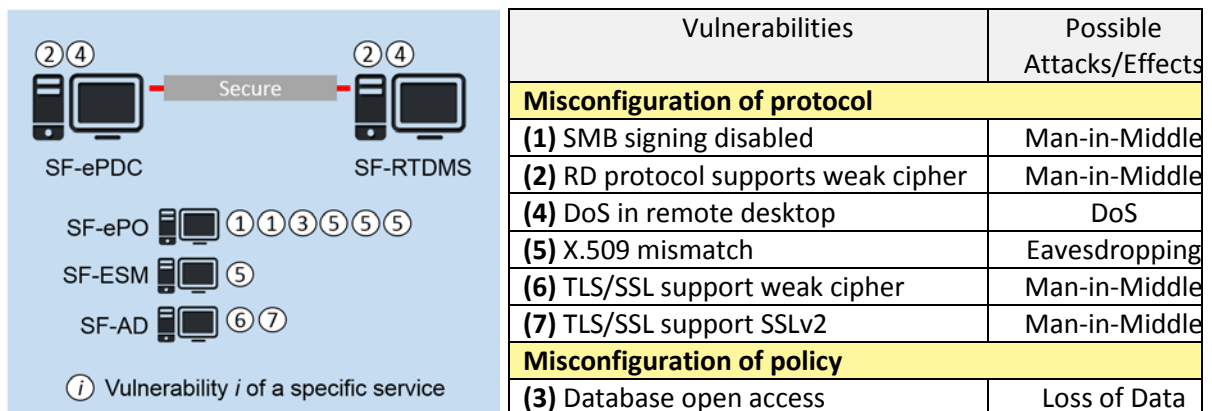


Figure 45. Vulnerabilities in the TTU Security Fabric-enabled Network

Using the same notation as in the previous figure, the diagram on the left above indicates the location of the vulnerability. For example, testing revealed six vulnerabilities within the SF-ePO: three of type (5), X.509 mismatch; two of type (1), SMB signing disabled; and one of type (3), Database open access. The table on the right shows that the SF-enabled system has four types of vulnerabilities (4-7) that were not found in the non-SF-enabled network. The DoS in remote desktop is the most critical one as it could lead to a Denial of Service (DoS) attack so it required immediate attention but was easily mitigated by updating the version of remote desktop. The rest of the vulnerabilities are considered severe, but they are often harder to exploit and may not provide the same access to the affected systems.

As explained previously, the diagram indicates that the total number of vulnerabilities of the Security Fabric-enabled network is 13 which is more than the six found in the non-SF environment. This higher count is because the SF-enabled system includes additional components (e.g., SF-ePO, SF-ESM and SF-AD) that are also vulnerable. However, a number of points here are important to note:

- All but one of the vulnerabilities identified were caused by misconfiguration or older, less secure versions of software, so these were easily fixed.
- The connection between the two endpoints, ePDC and RTDMS, in the SF-enabled version actually had less vulnerability than the non-SF-enabled version, so it is providing a higher level of protection which was the objective.
- All of the identified vulnerabilities have been mitigated, and Intel/McAfee has updated its support guides to ensure that these potential vulnerabilities are identified and mitigated during installation of the SF.

The details for each of the vulnerabilities, and the scripts used to identify the vulnerabilities during penetration testing are presented in Appendix B of Attachment 16.

4.5.2 Metrics and Benefits Analysis

The Security Fabric was designed to address all seven of the NIST-IR 7628 security guidelines. The approach integrated the best of breed commercial technologies to operate in parallel with the synchrophasor applications. The approach was designed to be compatible with standard IT processes, and was designed for interoperability. The approach leveraged the silicon capability of server hardware in secure identification/authentication, together with partitioned servers and virtual machines running the application software. While the Security Fabric has some “knowledge” about the application it is protecting (such as which devices the application is expected to communicate with), the application does not need to be modified in order to be protected by the Security Fabric. In achieving this design goal, the Security Fabric is extensible to cover other critical operating functions that demand cyber security protection. The specific metrics to be captured and reported are unique to this project, and were defined during laboratory testing and device integration.

4.5.2.1 Operational Benefits

The benefit of the Security Fabric approach is to provide data transport security for an IP connection between two locations without the need to make special modifications to the applications that will be communicating across the connection, and without materially affecting the performance between the applications. Success in this demonstration enhances the deployment and use of synchrophasor applications inside the operating control rooms of independent system operators by eliminating concerns about the security of the data being delivered from the TDSPs.

4.5.2.2 Build Metrics and Benefits

Associated hardware and software costs for this project component were born by Intel/McAfee and were not contributed as cost share in order to avoid any continuing ownership or licensing obligations.

4.5.2.3 Impact Metrics and Benefits

The application of the Security Fabric framework does not result in any energy savings or environmental pollution reductions so no impact metrics are defined.

4.5.3 Stakeholder Feedback

During the development, installation, and testing of the Security Fabric products, there was no direct stakeholder involvement or feedback. Since that time, ERCOT and a number of TDSPs have expressed interest in the testing results, and in learning more about how this technology can be applied to other systems requiring significant cyber security protection schemes.

4.6 Conclusions

This section covers projections of demonstration and commercial scale system performance as well as lessons learned and best practices.

4.6.1 Projections of Demonstration and Commercial Scale System Performance

As noted above, the benefit of the Security Fabric approach is to provide data transport security for an IP connection between two locations without the need to make special modifications to the applications that will be communicating across the connection, and without materially affecting the performance of the applications. In this project, the synchrophasor data flow between a simulated TDSP PDC and an ERCOT PDC was secured by the Security Fabric. A natural extension of this approach will be to encapsulate the synchrophasor data flow across the TDSP network between the PMU at the substation and the PDC. Extension of the Security Fabric approach to other critical infrastructure applications is also anticipated.

4.6.2 Lessons Learned and Best Practices

Almost immediately upon connecting the Security Fabric systems to the TTU network in December, 2013, significant traffic originating from China and Eastern Europe was identified as attempting to access the resources of the protected devices. In general, the attacks observed consisted of generic secure shell brute force exploit attempts. These are common on the Internet and are generally automated. Sifting through the logs indicated that all the brute force attempts failed to authenticate and only whitelist IP addresses have been successful thus far in accessing the systems. There has been evidence in the logs that someone tried to access one of the protected services, but again, the systems were successful in denying those attempts.

During security fabric testing performed at TTU, the Team first performed an initial exploration of every accessible node in the overall network. During this testing, the Team discovered a vulnerability in an SEL-734 PMU that allowed one to bypass authentication in the PMU's software controller. This type of vulnerability could lead to denial of service, and loss of data confidentiality and integrity. Such a vulnerability is new and not publically documented before. The vulnerability is due to the implementation flaws of the software controller of specific-domain hardware. Even a powerful scanner was not able to detect this vulnerability, so it needs to be addressed when these types of PMUs are deployed in external environments that may not be secure.

With respect to best practices, upon discovery of potentially nefarious network activity, all passwords in all systems should be reviewed for their strength, and where necessary, the password strength increased. Also, a password rotation policy should be standard procedure to regularly change passwords. Furthermore, in some instances, such as the synchrophasor network, the practice of allowing only whitelisted IP addresses to access the system can reduce the vulnerability to successful intrusion.

Another recommendation is to employ a vulnerability scanner on the networks, and use the results to mitigate potential vulnerabilities. Specific to the Security Fabric framework, it is imperative to perform this function and remove all vulnerabilities prior to implementation because this secure framework does not allow for remote updates.

5. RESIDENTIAL CIRCUIT MONITORING

5.1 Introduction

Residential rooftop solar PV distributed generation is logically a good complement to wind resources on the grid since the solar contribution comes during the day while the bulk of the wind contribution (west Texas wind) is concentrated in the late evening and night-time hours. Concentrations of rooftop solar PV generation, however, bring their own challenges to the grid operator, and these are likely to be present during the shoulder months (March – June, and October – November) when the system demand is low, the sun is shining brightly, and west Texas wind is most abundant. A somewhat similar challenge occurs for high wind contributions during these same months. Taken together, solar plus wind concentrations during the shoulder months may present a number of challenges both to the distribution utilities and the grid operator. This circuit monitoring component of the CCET project was intended to investigate and inform the stakeholders in the ERCOT market about the key challenges of concentrated rooftop solar PV distributed generation on the distribution system, and identify areas where research is needed to more fully understand and deal with these challenges.

One of the challenge areas resulting from increasing market penetration of residential rooftop solar PV distributed generation is variations in electrical output and performance parameters on residential distribution circuits and behind the meter, such as excursions in net load on the utility, power factor, voltage, and harmonic distortion effects. As reported herein, these issues arise from multiple interacting and dynamic factors, including variable distributed solar generation cancelling utility kW load, increasing harmonic current drawn by modern home appliances, and others.

This project component included residential circuit monitoring at two levels: 1) on 35 kV underground residential distribution (URD) circuits in Houston, Texas, and 2) at the residential customer point of connection in Austin, Texas. CenterPoint Energy provided data collection and analysis for the 35kV monitoring effort in Houston, while Pecan Street Inc. provided data collection and analysis for the residential service monitoring effort in Austin. The following subsections provide detailed discussions of conditions at both locations.

5.1.1 Power Quality on 35 kV Underground Residential Circuits

CenterPoint Energy led an effort to examine correlations between distributed solar PV generation and distributed circuit power quality impacts by measuring and comparing electrical parameters on a total of six 35 kV circuits in two neighborhoods in an area north of Houston, Texas, one with a moderate penetration of solar PV and the other with virtually no solar.

The Harmony neighborhood (formerly Discovery at Spring Trails) has a high concentration of new homes (built in 2008 and after) with very efficient building features, energy efficient appliances, small (average

1.1 kW) rooftop or trestle-mounted solar PV systems, and HEM systems for monitoring electric and water use. The second neighborhood, Legends Ranch, includes similar type homes, but they are built without enhanced energy efficiency features and are not equipped with rooftop solar PV, energy efficient appliances or HEM systems, i.e. more typical of newer Texas neighborhoods.

5.1.2 Power Quality at the Residential Service Point of Connection

A total of seven homes in the Mueller neighborhood in Austin, Texas, were monitored for power quality parameters, solar PV production and net load served by the utility. The Mueller area is a relatively new, “green built” neighborhood with efficient home construction, energy efficient appliances, a high concentration of installed grid-tied solar PV and plug-in electric vehicles.

In the Mueller neighborhood, approximately 210 out of the 650 homes on the dedicated distribution circuit have solar PV, so the neighborhood has approximately 30% solar market penetration. The average installed system nameplate DC capacity is 5.24 kW for each home having solar PV.

The homes monitored for this study were selected because they share two 50 kilovolt-amperes (kVA) distribution transformers, and because they are participating in Pecan Street research trials, whereby the residents have agreed to allow their energy usage to be monitored. The two transformers for the selected homes were monitored allowing the data to be compared with measurements on individual homes. The transformers were loaded as follows:

- Transformer 7007 has 10 connected houses, 3 with solar (30%); 18,800 total square feet
- Transformer 7006 has 8 Houses, 5 with solar (63%); 14,200 total square feet

5.2 Anticipated Benefits

Ultimately, the expected benefits of the CCET demonstration project would derive from a more reliable electric grid that could facilitate effective management of and responses to increased wind and solar resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage. Customers would be empowered to effectively and reliably manage their peak demand, therefore resulting in reduced customer electricity costs, reduced system-wide capacity needs, reduced electrical losses, and reduced environmental impacts. Specific smart grid benefits that are potentially supported by the residential circuit monitoring component center around informing utilities about potential PV impacts to power quality and reliability. The following two sub-sections provide more detail on these potential benefits.

5.2.1 Power Quality on 35 kV URD Circuits

Other research has been being undertaken by organizations like EPRI, the California electric utilities, the Smart Inverter Working Group, equipment manufacturers and others in the U.S. and abroad to measure and mitigate power quality impacts coincident with residential solar PV installations, ramp rates from

varying solar production, response of solar inverters to system transients, and other issues. This CCET effort was focused on collecting electrical data, performing graphical analyses and making observations about how power quality varies with daily and seasonal swings in load and PV production in targeted installations in Houston. This effort did not extend to defining or testing solutions.

Specifically, this effort focused on analysis to compare and correlate solar PV generation to circuit parameters by measuring and calculating total harmonic current and voltage distortion (THDi and THDv, respectively), power factor (PF) and voltage at the distribution level as 5 minute averages in two neighborhoods, one with a small to moderate installed capacity of solar PV and one without solar PV generation. The observations noted from this effort, combined with findings in the other “Residential Service Point of Connection” element of this investigation (below), will help utilities anticipate the occurrence and assess the significance of potential power quality impacts in communities with a range of concentrations of solar PV and in making design and/or operational adjustments to mitigate such impacts.

5.2.2 Power Quality at the Residential Service Point of Connection

In general, this investigation provided an opportunity to develop a more complete understanding of the interaction of solar PV with the distribution infrastructure. The root cause of any observed issues should be traceable to events occurring within the residential load, residential solar PV operation, or the utility distribution service. Information gained on the secondary side of the service transformer can be compared with measurements at the distribution circuit level (even a different circuit) if the load type and distribution between the two circuits are similar.

Previous studies (Grady, *et al.*, 1993) have shown residential power factor to be .9 to .95 lagging, and typically of low distortion.⁶ The addition of solar PV and the proliferation of switch mode power supplies are changing these characteristics.

The benefit to the utilities is an updated and more accurate characterization of the residential electrical load at the point of connection.

5.3 Interactions with Key Stakeholders

The primary stakeholders in the 35kV URD circuit monitoring study were the electric utilities who have an obligation to deliver electricity with power quality meeting industry standards. Analysis observations and conclusions have been discussed among the electric utilities on the project team.

Stakeholders interested in the monitoring at the service point of connection included utilities, manufacturers of distribution equipment and consumer appliances, and the residential customers

⁶ Grady, W. Mack and Robert J. Gilleskie, Harmonics and How They Relate to Power Factor, EPRI Power Quality Issues & Opportunities Conference (PQA'93), San Diego, CA, November 1993.

themselves. The utilities on the team have analyzed and discussed the data and findings. Customer participants in this study have been provided feedback on the findings. Consumer electronics companies on Pecan Street research teams, including LG Electronics, Schneider Electric and others, will be made aware of the impact of their devices on the grid, as observed in this investigation. Analyses and observations also will be shared with distribution equipment manufacturers.

5.4 Technical Approach

This section provides a detailed description of the Project Plan and the Data Collection and Benefits Analysis.

5.4.1 Project Plan

The goal of the monitoring and analysis in Houston and in Austin was to identify, and quantify to the extent possible, the effects of solar PV distributed generation on power quality in neighborhoods with a significant concentration of installed distributed solar PV. Implementation steps included:

- Identification of the presence and potential drivers of power quality impacts.
- Identification of data needed for analysis.
- Identification, procurement and installation of data monitoring and collection equipment.
- Development of a system to collect and store data.
- Performance of data analysis.
- Establishment and characterization of instructive correlations among power quality, loads and solar production.
- Differentiation of effects in neighborhoods with and without significant distributed solar PV installed capacity.
- Suggestion of areas for further study to address technical challenges and maintain quality of service.

5.4.2 Data Collection and Benefits Analysis

Details are provided below for the 35 kV URD circuit monitoring and residential service connection monitoring and analysis efforts.

5.4.2.1 Power Quality Impacts from Solar PV on 35 kV URD Circuits

For this effort, CenterPoint monitored three 35 kV URD circuits in each of two neighborhoods. The following table provides some information on the circuits, including respective universal circuit monitor (UCM) identification numbers, home counts, average square footage, and the amount of installed solar PV for the Harmony and Legends Ranch neighborhoods:

Table 15. Basic Home Information for Both Neighborhoods

Legends Ranch	# Homes	Average Square Feet	Solar kW Installed ⁷	Average Solar kW per Home ⁸
UCM4	43	3,272		
UCM5	44	3,740		
UCM6	94	1,156		
Harmony				
UCM2	60	3,361	66	1.10
UCM3	26	4,418	30	1.15
UCM7	40	2,739	46	1.15
			142	1.13

As represented above, installed capacity was 142 kW in 126 homes, or slightly greater than an average of 1 kW per home, for the three monitored circuits in Harmony. The largest known single installation was reported to be 6 kW.

It should also be noted that there were initially some technical issues related to data communications, streaming, storage and integrity with the original installed monitors, but all of these issues were resolved by the installation of new monitoring equipment in January 2014. From that point forward, the following parameters were captured by SATEC PM174 PQ Monitors.

- Power Factor (PF)
- Frequency
- Voltage
- Current
- THDv
- THDi
- Real Power
- Reactive Power
- Apparent Power

The data was collected as average values for each five-minute interval. The analysis also utilized the following weather data for the site: hourly temperature, relative humidity, and direct and horizontal solar irradiance (used as a proxy for solar PV generation, which was not directly metered).

The analysis looked for correlations between THDi, THDv, voltage, solar PV output, and URD circuit load (kW). Power quality impacts at Legends Ranch (no solar) and Harmony were compared in magnitude

⁷ Note: installed solar PV capacity in Legends Ranch is unknown, but is reported to be minimal or zero.

⁸ Ibid.

and seasonality. Power quality performance was compared to findings from monitoring at the residential point of connection (below).

5.4.2.2 Power Quality Impacts from Solar PV at the Residential Service Point of Connection

The basic steps applied for the service point of connection monitoring were identical to the 35 kV URD circuit monitoring approach—identify data needed, install data collection equipment and analyze the data.

Seven homes were selected for monitoring in Austin. These homes were served by two different transformers, and one of the transformer clusters had high solar PV penetration. One home in the Houston Harmony neighborhood was also monitored to explore similarities and differences between the 35 kV URD circuit data and the customer service line data and to examine whether the solar inverter was contributing to high THDi.

The Team connected a Power Standards Lab PQube power quality and energy monitor/analyzer at the meter and installed E-Gauge sub-meter data collectors at the main power panel to collect and calculate the following:

PQube at Customer meter:

- Volts (leg-neutral, 12-cycle average)
- Volts (leg-leg, 12-cycle average)
- Amps (12-cycle average)
- Ambient Temperature and Relative Humidity (%)
- Frequency
- Harmonic voltage distortion (% THD_v)
- Harmonic current distortion (% THD_i)
- Flicker
- kW
- kVA
- kVA reactive (kVAR)
- Total power factor (PF)

E-Gauge sub-metering:

- Net usage
- Actual usage
- PV generation
- Heating, ventilation, and air conditioning (HVAC) condensing unit
- HVAC air handler
- Electric vehicle charging

- Clothes washer and dryer
- Dishwasher
- Disposal
- Kitchen appliances 1
- Kitchen appliances 2
- Oven 1
- Oven 2
- Refrigerator 1
- Refrigerator 2
- Microwave
- Lights 1
- Lights 2
- Sub-panel

Graphical analysis was most useful in finding and reporting correlations between fundamental current, distortion current from home appliances, THDi, net utility load (net solar production) and power factor.

5.5 Performance Results

5.5.1 Operation of Smart Grid Technologies and Systems

From this point forward in this document, reporting of results combine the Houston 35 kV URD circuit monitoring and the Austin residential point of connection monitoring efforts, as they work in tandem to explain the observations.

The five minute interval data for the 35 kV URD circuit was graphed and then statistically analyzed to quantify the degree of correlation between concurrent values of solar production and several other parameters (power factor, voltage, THDi and THDv).

$$\text{THDi} = \text{distortion current (amps)} / \text{fundamental current (+ or -)}$$

In the equation above, two things can increase THDi: 1) increasing distortion current (such as by adding advanced home electronics with switch-mode power supplies, compact florescent lamps, low quality solar PV inverters, high efficiency HVAC equipment with variable speed fans and compressors, etc.), and 2) reduced fundamental current drawn from the utility, such as when solar kW production cancels most (or all) of the demand for utility-supplied fundamental current. When the denominator of the THDi equation approaches zero (for example, when solar production offsets the entire home load, which readily happens in cool weather), the THDi value approaches infinity.

Data was collected and analyzed from February 2014 through October 2014, which allowed examination of seasonal differences in circuit parameters for both Harmony and Legends Ranch. Additionally, monitoring of a Harmony house by Pecan Street at the service connection that was also on a circuit

monitored by CenterPoint provided the opportunity to cross-reference observations between Houston and Austin, and to postulate drivers for circuit behavior, as presented below.

The following are the primary observations from these investigations:

1. In both Harmony and Legends Ranch, THDi varies throughout the day, operating within a seasonal range.

In Harmony, cooler months such as February to March (graphed below) are characterized by light overall utility circuit loading due to seasonally reduced HVAC demand, concurrent with strong solar PV production that offsets most of the utility circuit load on cool sunny days, driving the THDi to higher values.

Observations – Cool Weather THDi

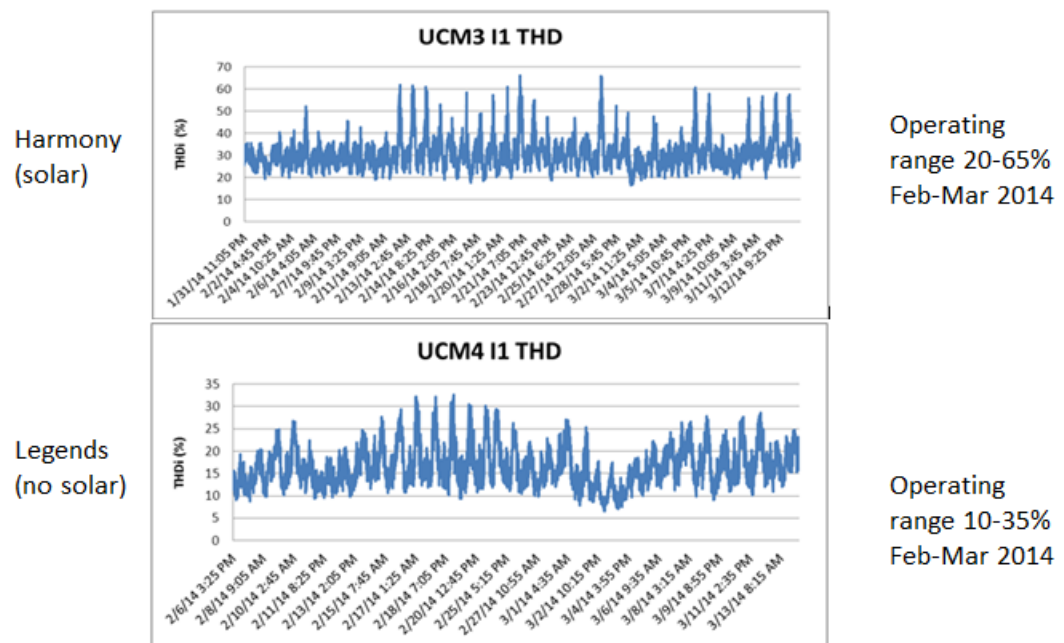


Figure 46. Observations for Cool Weather THDi

In contrast as shown below, hot weather during July to August, which calls for longer HVAC run times, causes the utility-supplied kW to far exceed the available solar production as shown below. This “washes out” the harmonic currents, keeping THDi lower than in cooler weather (shown above).

Observations – Hot Weather THDi

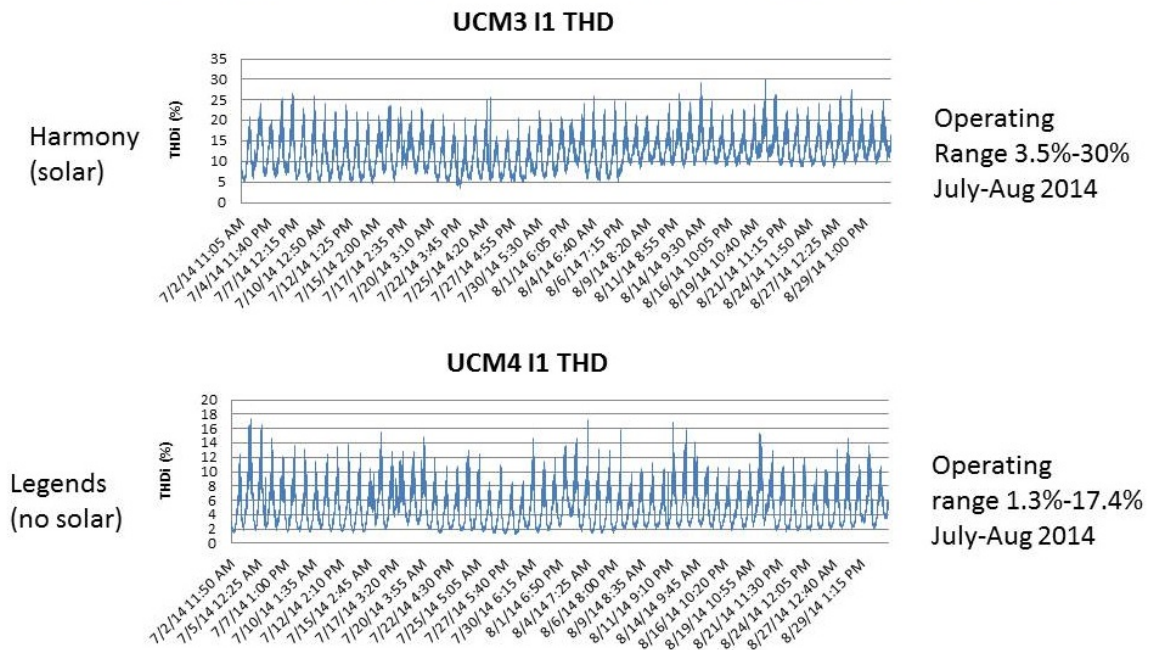


Figure 47. Observations for Hot Weather THDi

It is important to note that the daily cycles in THDi (above) are not the result of large magnitude variations in harmonic distortion current inside the home. The harmonic distortion current measured at the home consistently varies in a range of only 1 to 5 amps for each 120v leg, as shown on the next graph from a home in Harmony which has a 1 kW PV system installed.

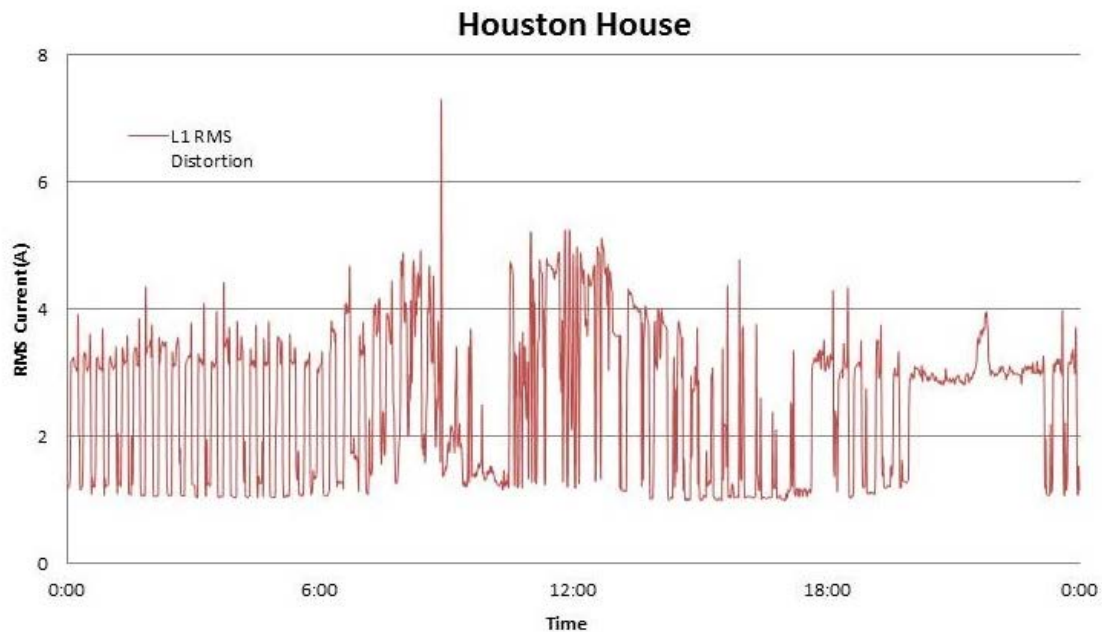


Figure 48. Harmonic Distortion Current Measured at a Houston Harmony Home

The cancellation of utility load (fundamental current) by solar production, as in the day-time hours on the next graph, can cause the THDi in a given home to spike to several hundred percent (blue line) and fluctuate based on HVAC cycling and cloud cover (rapidly varying solar output). The green line below is grid current delivered by the utility to serve the net load. Note that when the green line approaches zero, the blue line (THDi) spikes.

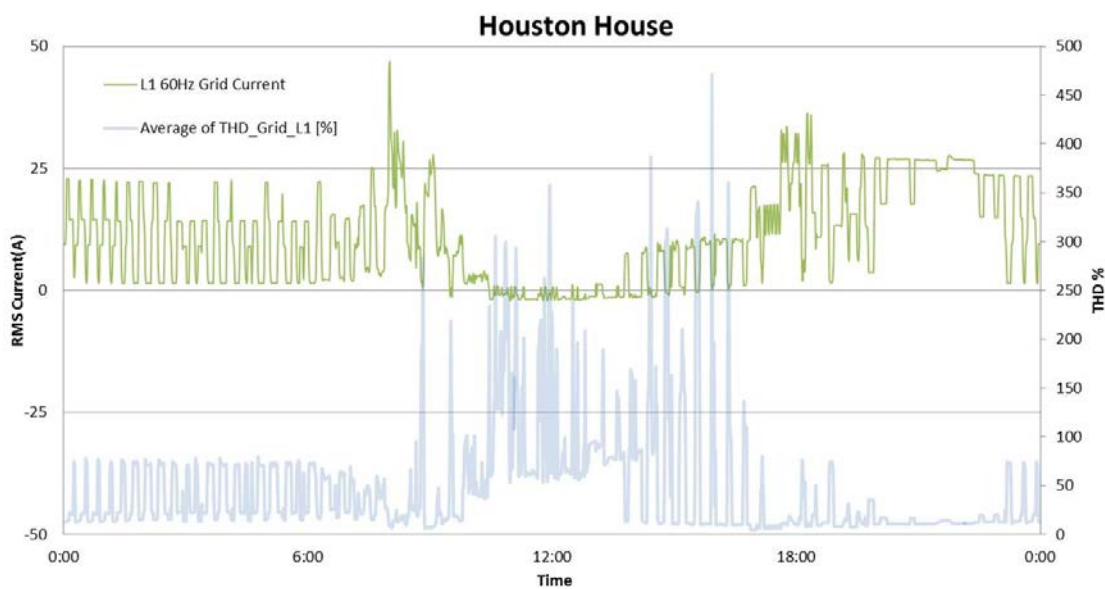


Figure 49. Relationship of THDi to Grid-Supplied Fundamental Current

Thus, elevated THDi is not a result of increased harmonic current—the numerator of the THDi equation – but rather a result of the denominator (fundamental current from the utility) approaching zero at net-zero load (when solar PV production balances home load). This is a dynamic balance that cycles throughout the course of a 24-hour period on a given home, depending on solar production (as affected by cloud cover) and home load (e.g. cycling HVAC and other appliance loads).

At the distribution circuit level, the next graph of UCM2 in Harmony illustrates the relative movement of THDi, solar production and net utility load throughout a sunny day in April, when solar production is able to cancel all but 20 kW of the entire utility load (60 homes).

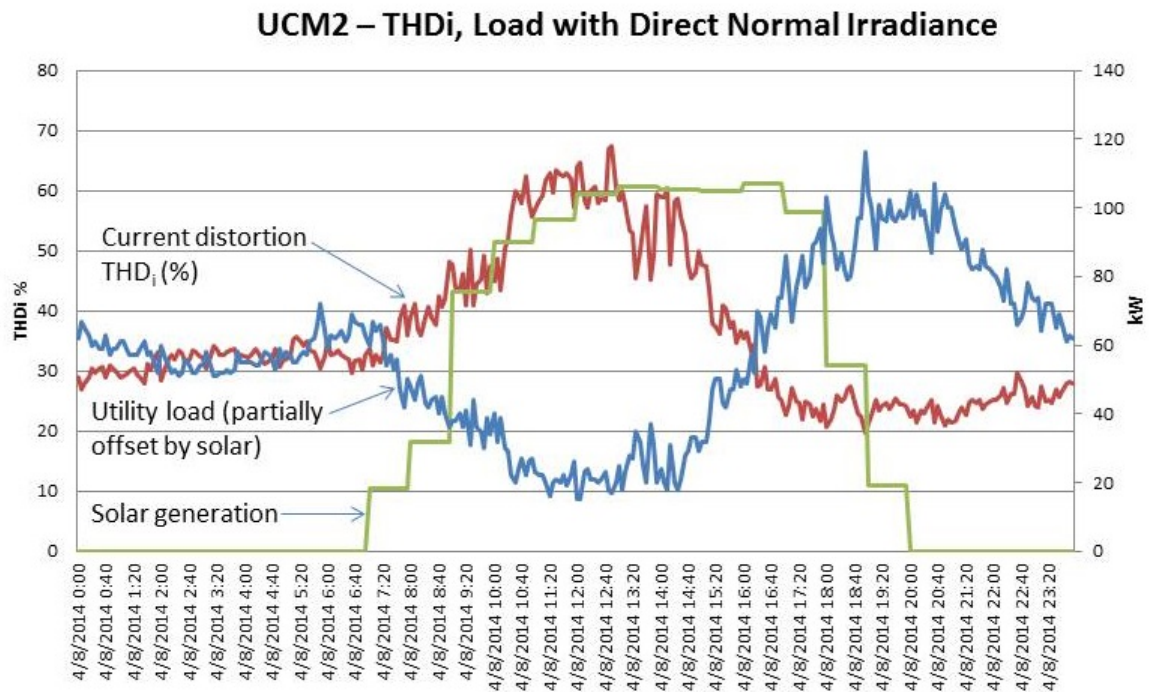


Figure 50. Movement of Harmony THDi, Solar Irradiance and Utility Load on a Sunny April Day

In contrast, as shown below, with the absence of the solar PV influence, THDi on the Legends Ranch circuits varies throughout the day, and is lower in the daytime, when kW load on the circuit is higher than at night.

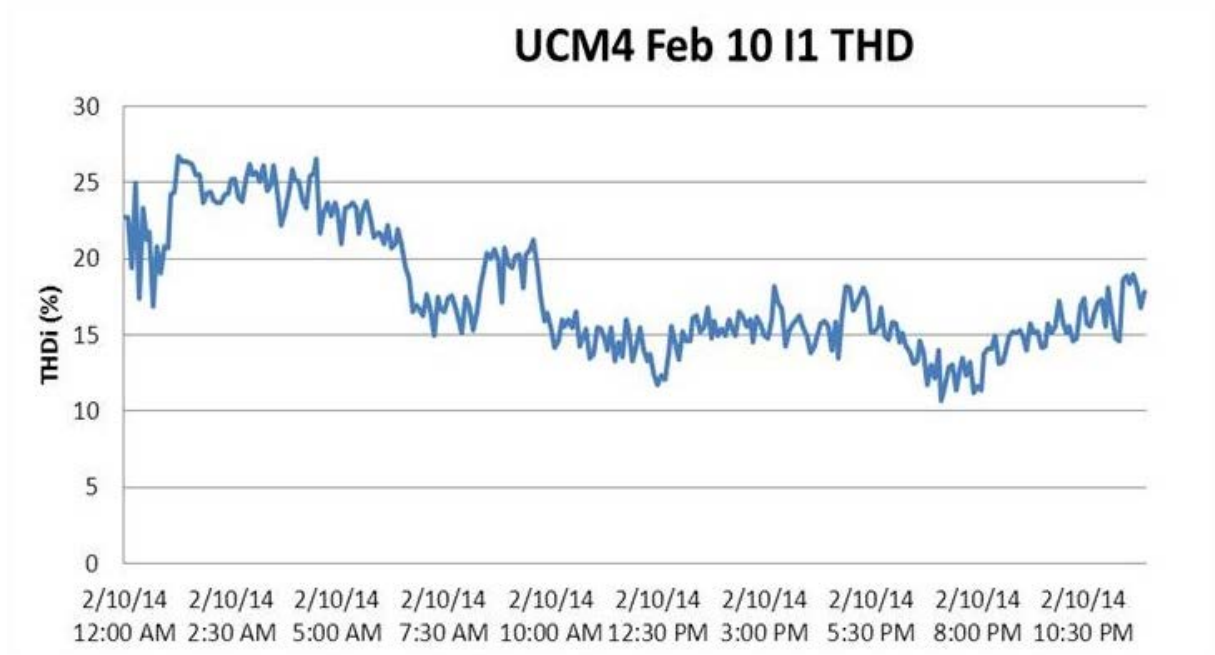


Figure 51. THDi Movement in Legends Ranch

Based on the results shown above, the Team drew the following conclusions:

1. THDi at distribution voltage:

- Reflects a dynamic composite of all the varying loading conditions for each connected service transformer
- Is unique per circuit
- Varies throughout the day
- Depends primarily on the weather (HVAC load, i.e. seasonal) and the amount of solar PV generation offsetting the utility load

The solar inverter does not add significant distortion to the grid; rather the presence of solar production enables the steady distortion from other devices to appear higher relative to utility loads.

2. In communities with significant solar PV installed, there is an inverse and seasonal relationship between power factor (PF) and solar production at distribution voltage.

The resulting daily swings in PF are depicted in the Harmony UCM2 graph below. PF dips as solar PV offsets circuit load. This is an April day (not a very high HVAC load).

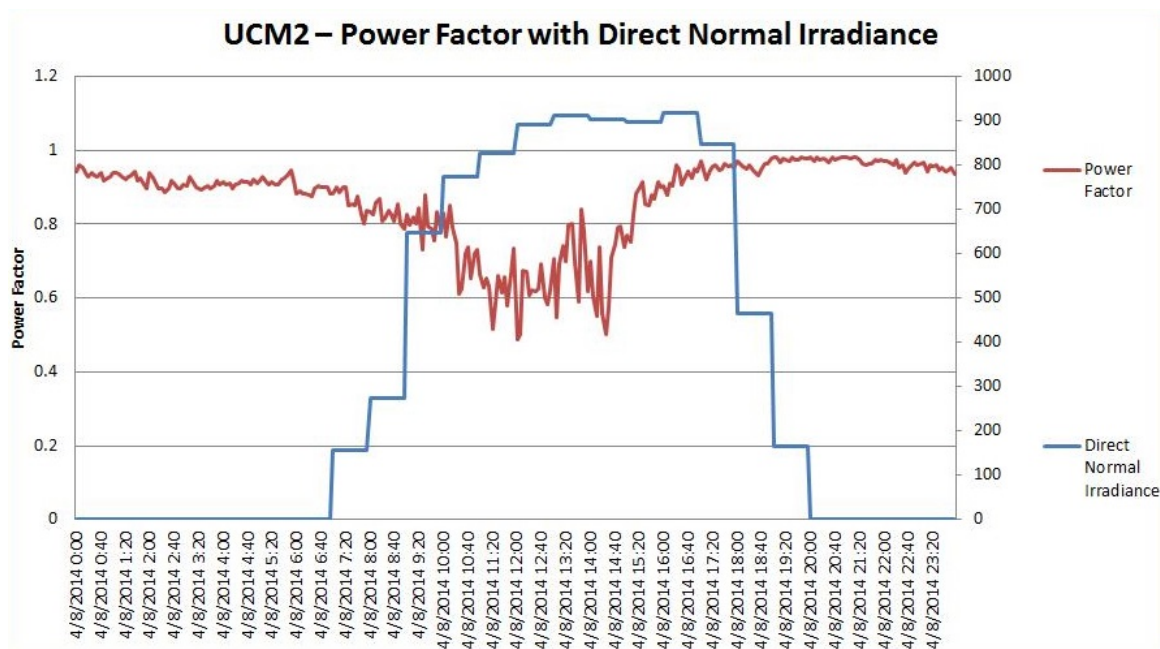


Figure 52. Daily Swings in PF During April 2014

Seasonal graphs of PF follow for Harmony and Legends Ranch, showing much milder daily PF swings in hotter weather, as with the seasonal swings of THDi in the previous graphs, due to the higher load on the circuit.

Observations – Cool Weather PF

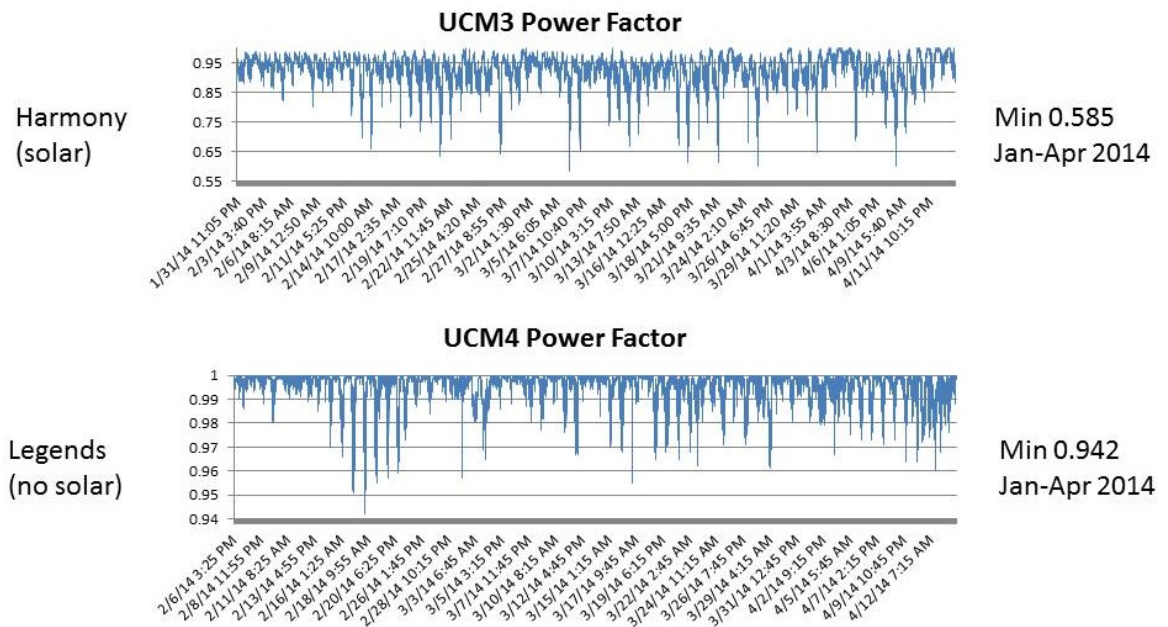


Figure 53. Cool Weather PF Observations

Observations – Hot Weather PF

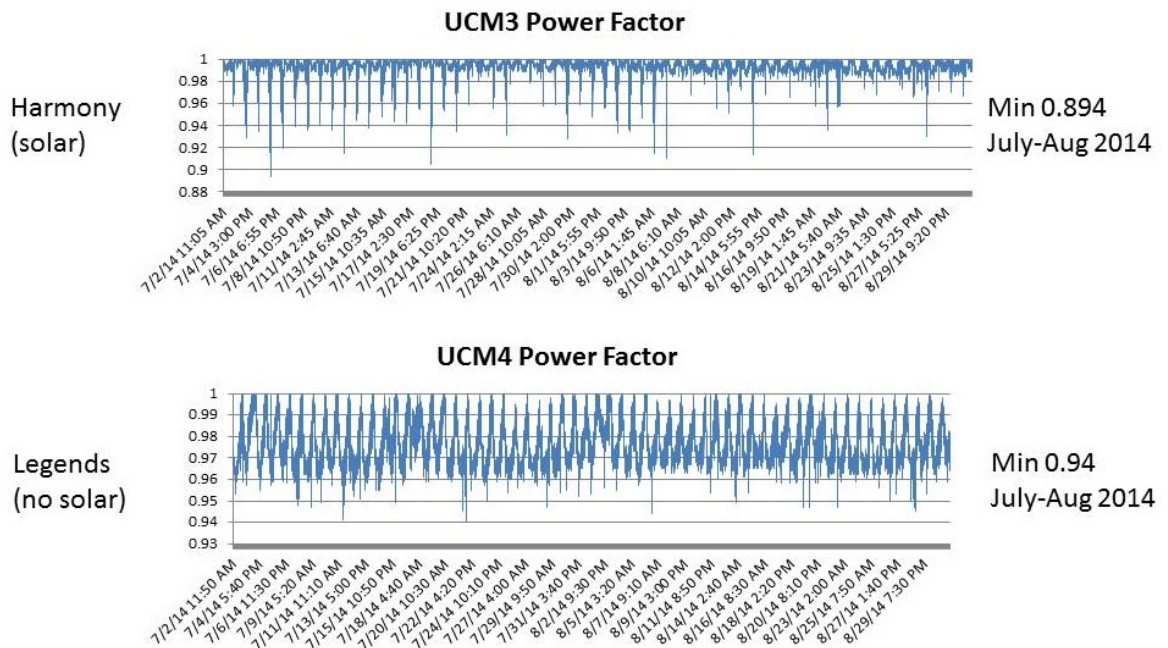


Figure 54. Hot Weather PF Observations

3. To determine whether the THDi and PF behavior observed on the 35 kV URD circuits in Houston was unique to that neighborhood, similar monitoring and analysis was performed on the homes in the Mueller neighborhood in Austin. The same type of THDi and PF behavior as reported for the home in Houston was observed in the Austin residential monitoring. For the Austin houses with larger solar arrays, longer periods of fundamental load cancelation were observed, though this will vary by day, weather, array size and home electricity usage. This behavior will likely be seen at some point during the year by any house where solar PV production can offset an appreciable portion of the total home kW load.

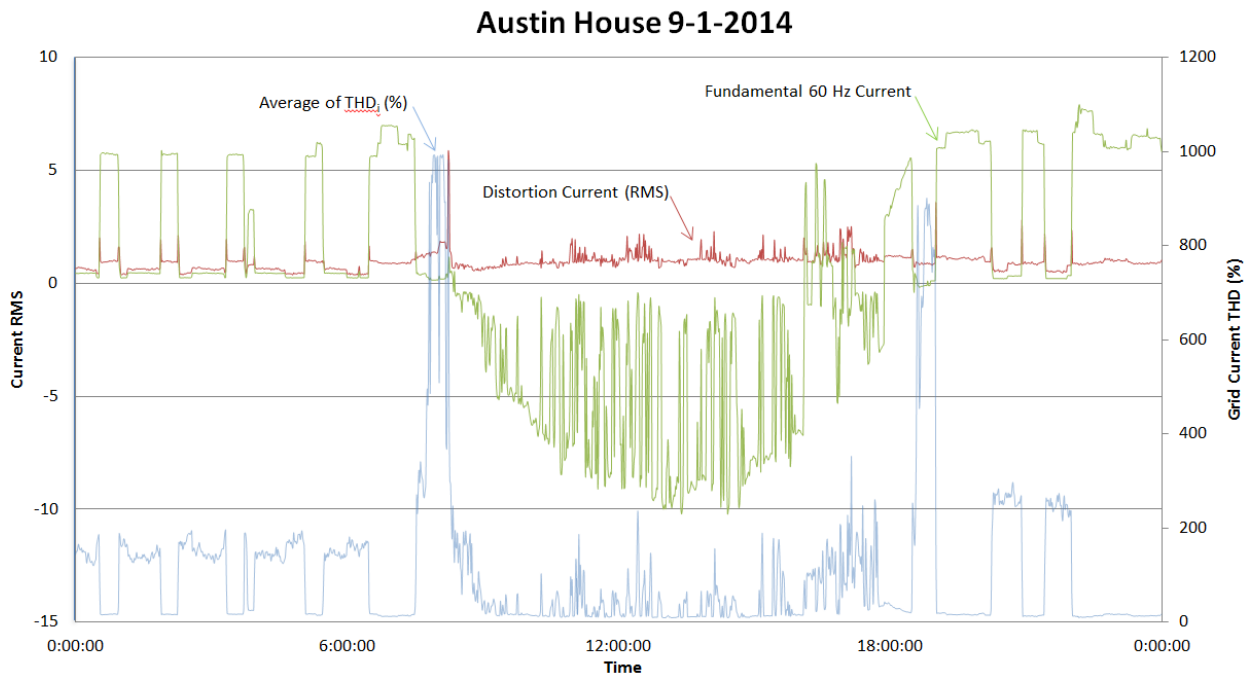


Figure 55. THDi Behavior for Home with Larger Solar PV System

In the figure above, the distortion current stays approximately the same during the entire day with variations that can be attributed to device usage, not solar production. During the daytime, solar capacity exceeds home load much of the time, driving the net utility load (green line) negative, with oscillations due to cloud cover and home appliance operation (generally HVAC). During periods when the net utility load is canceled and approaches zero for minutes to an hour at a time, the THDi spikes are very high, up to 1000%. Again, this is due to the very low amounts of fundamental current supplied by the grid, not the solar production adding distortion to the grid.

4. Referring to the graph below, PF at the service connection for a house with several kW of solar PV installed capacity varies from 1 to -1 during the day, changing rapidly based on load and solar production. Pecan Street analyzed power factor (PF) at the individual residential service connection. As mentioned above in #2, the Harmony circuit showed a general drop in PF during solar production hours. With PF being defined as real power divided by total apparent power,

during periods of solar production the real power delivered to a given house from the utility can be negative (as in #3 above), with real power flow to the utility and away from the load.

In the next graph, an example day for a single home in the Mueller neighborhood shows a large number of negative/positive swings in utility load and PF, and periods of operation at very low PF.

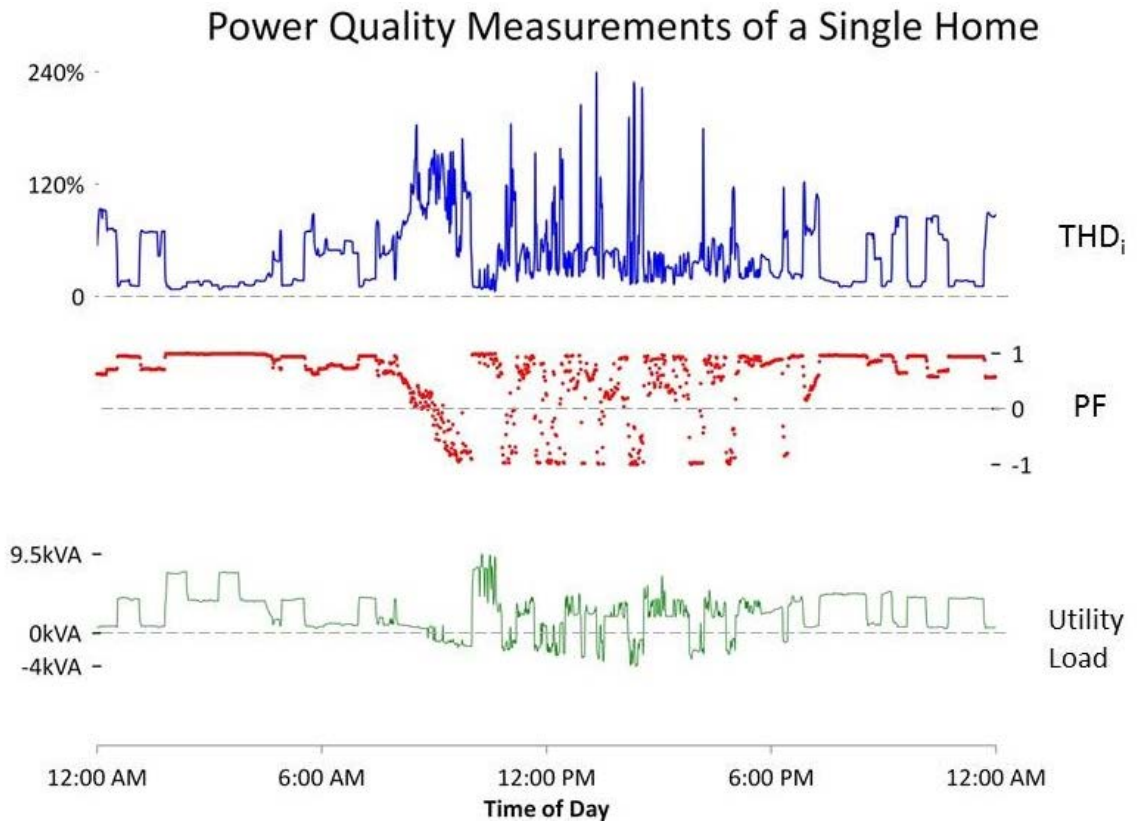


Figure 56. Power Quality Measurements for Single Home in Austin

It should be noted that this home also shows large swings of THDi as the total load hovers around 0 kVA. Note also that the power factor starts dropping as the solar generation comes online. In this case, the utility load is also significantly more dynamic than might normally be seen.

Referring to the figure above, notice that this home, which has an electric vehicle and solar PV, went from being a 2 kVA net generator at roughly 10 a.m. to a 9.5 kVA net load in less than one sample period (one minute), indicating a rate of change of over 11.5 kVA/minute. This is a substantial and rapid increase in power consumption not typically seen in traditional residential loads. However, in homes with large solar PV, HVAC and PEV charging capacities/loads, such rapid changes will now be fairly frequent.

In addition, the power factor at this home remains at levels well below 0.9 total PF (which measures displacement and distortion based power factor) for long periods of time. The impact on the distribution circuit will be determined by the resolved combination of localized real and reactive power at each service transformer. The utility typically controls power factor on their distribution circuits by adding switched capacitor banks to cancel out the inductance of transmission line or motor loads.

There are existing guidelines for current harmonic distortion on the grid, typically expressed as Total Demand Distortion (TDD). TDD is the ratio of instantaneous distortion current to total available supply current, as compared to THDi, which is the ratio of distortion current to instantaneous fundamental current. Inductor/capacitor trap circuits are used to mitigate harmonics. The TDD of the house above, which is representative of all houses in the study, is not particularly high, less than 5%. However, IEEE-519, the standard for total distortion limits in distribution circuit design, assumes that the effective source impedance of the distribution grid is loaded with some amount of resistance effectively damping the possibility of oscillation and loss of power factor control. It is unknown to the authors the impact of removing the resistive load (through the generation of real power from solar production) on the stability of what is effectively an RLC (resistive-inductive-capacitive) circuit. The removal of the real resistance R by high levels of solar PV generation may induce a high Quality Factor LC resonant circuit, resulting in potential instability on individual loops or circuits.

The house above goes from being a net load on the utility to being a net generator at least 14 times in one day. Utility scale capacitor banks are switched according to utility operational practice, which is likely not fast enough to handle the change rates seen here.

In addition the capacitor banks can only cancel lagging VARS and cannot cancel leading VARS or VARS due to distortion currents. Switch mode power supplies can commonly draw over 100% THD in the current waveform, and therefore the utilities may increasingly find themselves in the situation where the power factor correction methods employed only correct lagging VARS while doing nothing for the periods of time when a residential load is leading or distorted. Separate measures must be taken to effectively cancel leading VARS or distortion VARS.

Smart inverters with reactive power support for both phase angle and distortion could be employed and would eliminate the issues, as would storage for voltage control, but right now there is little or no economic incentive for the residential customer to participate in these markets. Future research might identify the value of ancillary services provided by these technologies as well as potential methods to incentivize or compensate those who adopt and deploy them.

5. 35 kV URD circuit voltage was found to be weakly correlated to solar production in Houston. The Austin monitoring compared voltage at the service transformer to coincident voltage at a connected house with a sizable installed PV system. The next figure indicates that when the PV

is producing, the house service voltage exceeds the transformer voltage, i.e. the PV is supporting the utility-supplied voltage, as expected. The opposite is true at night, when the PV is not producing.

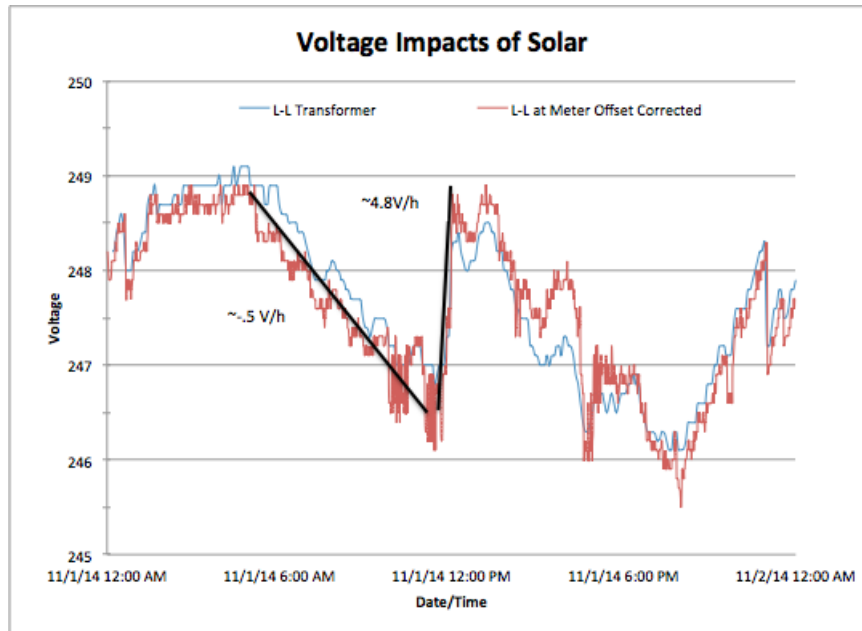


Figure 57. Comparative Voltage at Service Transformer and at House with Solar PV

In this figure, note that the voltage rate of change (ramp rate) varies from 0.5 volts per hour (without solar PV) to nearly ten times that rate when PV generation varies up and down due to cloud cover. Such extreme ramp rates, carried up to distribution voltage through the service transformer, could affect wide geographical areas served by a common circuit and affected by the same weather. On some circuits, it may be possible for the primary feeder voltage to experience voltage swings that confound normal methods of control such as substation tap changers. Anecdotal evidence shows that there are instances of solar inverters tripping offline during overvoltage conditions caused by the solar production. A look over the study period at the residential circuit level showed no such conditions, but there have been reports in areas outside the study area of longer service drops and feeder areas having inverters shutting down due to overvoltage operation at 252-255 V RMS or more.⁹

The typical methods of voltage control used by a utility may not have the ability to act quickly enough and often enough to counteract the voltage swings caused by the solar PV.¹⁰ If the equipment did operate fast enough to exert voltage control, the number of operating cycles required daily could easily

⁹ Requirement from the utility to allow the DG to be connected to the distribution system (e.g. +5% for 30 seconds). See PUCT Substantive Rule 25-212 for Texas.

<https://www.puc.texas.gov/agency/ruleslaws/subrules/electric/25.212/25.212ei.aspx>

¹⁰ The utility may require the owner of the PV generation equipment to mitigate any unacceptable voltage swings imposed on the distribution system.

be doubled, tripled or more if the observed behavior in the figure above is repeated frequently. Additionally, the graph shows that the voltage at the house can be approximately 1 volt greater than the transformer secondary voltage during times of high PV generation. In the observed neighborhood the distances from house service entrance to transformer is short, but in areas where the distances are larger (a block or more) this voltage rise figures to be greater. The typical utility voltage control solutions of multi-tap transformers or auto-transformers would be ineffective at counteracting this voltage change at the service point since they are located too far away from the load and the impedance between voltage control measures and load is too high. The figure appears to show a situation where voltage control is lost at a localized level.

5.5.2 Metrics and Benefits Analysis

5.5.2.1 Operational Benefits

1. This investigation presents observations, analysis and findings, and it poses questions and recommends further research.
2. A better understanding of drivers behind power quality on residential distribution circuits will greatly benefit utilities provide reliable and stable power to their residential customers via improved -
 - a. Planning for reliability
 - b. Monitoring and active mitigation if needed
 - c. Interconnection specifications
 - d. Distribution equipment specifications
3. From this investigation, a better understanding has been developed of power quality at the residential point of connection in the presence of PV production as relates to voltage, power factor and THDi.
4. A better understanding has been gained of the overall magnitude of distortion current draw by modern in-home devices and appliances. Examples are compact florescent lamps, switch mode power supplies, variable speed HVAC fans and compressors, and other electronic appliances. Such devices are themselves susceptible to power quality upsets.
5. From this investigation, a better understanding has been gained of the behavior of circuit parameters during times when PV production approximately equals the total load of the home. At such times, the utility is only “supplying” distortion current and reactive power, potentially resulting in impacts on stability, voltage control and equipment lifetime. These impacts include at the local level:
 - Increased power dissipation in fixed resistance devices due to higher voltage levels,
 - The potential for inverters to shut down in a safety response to overvoltage situations,
 - Locally poor power factor due to high density solar installations, and
 - Increased voltage rate of change, both up and down, leading to local voltage instability.
6. At the grid distribution level, increased switching of tap-changing and power factor correction devices could lead to reduced equipment lifetimes, insufficient voltage and power factor control, the need for more extensive monitoring and potentially faster acting power factor, and voltage control equipment.

5.5.2.2 Build Metrics and Benefits

Build metrics typically focus on equipment costs associated with and reported for a project. Although CenterPoint and Pecan Street invested significant resources for this effort, the only equipment costs included as cost share are related to the initial purchase of six UCMs, and that cost was \$89,128.

5.5.2.3 Impact Metrics and Benefits

Impact metrics typically define reductions in energy use and in environmental emissions as a result of technology implementations. These are not deemed relevant to this project.

5.5.3 Stakeholder Feedback

The primary stakeholders in the 35kV URD circuit monitoring were the electric utilities who have an obligation to deliver electricity with power quality meeting industry standards. Analysis observations and conclusions have been discussed among the electric utilities on the project team, and they will discuss them as well with their suppliers of distribution equipment.

Stakeholders in the monitoring at the service point of connection included the utilities, manufacturers of consumer appliances, and the residential customers themselves. The utilities on the team have analyzed and discussed the data and findings. Consumer electronics companies will be made aware of the impact of their devices on power quality, as observed in this investigation. Customer participants in this study will be provided feedback on the findings of this study, although no feedback is requested or expected from them.

5.6 Conclusions

5.6.1 Projections of Demonstration and Commercial Scale System Performance

7. The circuit monitoring and analysis performed in this study assists utility stakeholders and system planners in anticipating the degree of power quality impacts on residential distribution circuits as solar PV market penetration increases. This knowledge will help the utilities make adjustments to their system design, equipment specifications and operations in order to more accurately and efficiently predict, detect and address such conditions in order to maintain reliable electric service within the parameters of industry standards.
8. As the installed price of solar PV continues to decline, there will be continued increases in market penetration and average PV system capacity. Many homes already have several kW of installed solar PV. Such homes produce their own real power (kW), which at times cancels the real power supplied from the utility. To the extent this condition exists in a given neighborhood, the customer at times would be drawing only distortion current and reactive power from the utility and may be back feeding real power to the utility, leading to high THDi and potentially a sagging power factor on distribution circuits.

9. It was shown that measured THDi can be significantly higher on distribution circuits that feed high concentrations of homes with solar PV as compared to those serving homes with little or no PV. It is not clear that increased THDi is an immediate threat to reliability, but a review is suggested of the underlying assumptions of IEEE-519 regarding the damping effects of resistive loading. When the resistive loading is canceled by solar production, the impact of harmonics on grid stability may become an issue of concern that is not addressed by IEEE-519.
10. The PF variability on distribution circuits with high solar PV penetration will likely require enhanced monitoring and upsized (and fast acting) PF correction to meet operating standards. One means of mitigation could be VAR injection at the solar PV inverter. This gives rise to the discussion of compensation to the PV owner (customer) for any added equipment cost and for kW production forfeited in producing the injected VAR.
11. The localized voltage boost associated with residential PV production is additive at the service transformer and will propagate to distribution voltage. The resulting boost on the distribution circuit should not be problematic if the magnitude and rate of voltage change can be handled by existing utility voltage control equipment.
12. All houses on a common service transformer will experience boosted voltage from PV production on any one or more of the connected houses. The boosted voltage will vary at ramp rates impacted by fast moving cloud cover that affects PV production. Further study should be undertaken to determine the magnitude and potential impacts of limiting case service voltage fluctuations and the need for and means of local monitoring and mitigation.
13. This study confirmed with a high level of certainty that PV inverters conforming to industry standards do not add appreciably to voltage or current distortion.
14. Further investigation is suggested into the existing body of research (and need for additional research) on energy waste and appliance/device damage behind the meter due to compromised power quality under “zero net load” conditions.
15. Further investigation is suggested into the cost/benefit of stronger standards for harmonic distortion and power factor for residential devices to mitigate the increase in relative distortion (THD_i) in the presence of “zero net load” conditions occurring with solar PV generation. The current Energy Star standards for power factor on a CFL light bulb are only 0.5. Some homes in the subject Austin neighborhood have been found with 40-70 of these devices operating at any given time, yielding a total load of up to 1 kW at a PF of 0.6.
16. Expanded research is suggested on the need, feasibility, cost-benefit, and market implementation of distributed energy storage systems, placed either behind the meter or in the utility infrastructure. Such storage could allow for solar energy to be collected and stored for use throughout the day, minimizing power quality impacts by smoothing out rapid PV production ramps and if necessary mitigating zero net load instabilities.

5.6.2 Lessons Learned and Best Practices

The following outlines some lessons learned from this project:

- The 35 kV URD circuit monitors initially installed suffered from a lack of data storage capacity. Ensure that the monitoring equipment contains sufficient data storage, and provide a means for reliably transmitting that data.
- There were communication issues during the setup and testing phases (35 kV). Existing radio networks were easily interfered with from tall trees in the area so a decision was made to use fiber optic lines. First, there were issues with identifying the service providers since the project involved two different neighborhoods. Second, the fiber optic lines were continually being cut or dug up due to ongoing construction in the Harmony neighborhood. When the new URD circuit monitors were installed in January 2014, all communications were converted to a cellular network. So, it's important to carefully assess communications design requirements.
- The 35 kV URD circuit configuration of a neighborhood under construction changes over time, and therefore the URD circuits were redesigned as a new developer restarted the building of new homes in the Harmony neighborhood. This affected the ongoing reliability of the data being collected until the URD circuits were finalized, and the monitoring devices were relocated to better advantage.
- The project team worked with the developers and builders to get updated information on the homes as they were completed and connected to the 35 kV circuits. Certain information needed for analysis was obtainable from the developer and builders (square feet, information on installed HVAC). For studies such as this, it is important that these relationships be established early, and that the data capture efforts are established and standardized.
- Collecting energy usage, power quality and solar generation information across multiple hardware systems highlights the need for local data storage capacity. The systems used at the residential level used three different communications backhauls and there is difficulty in keeping them all online simultaneously over extended periods.
- Some power quality manufacturers have difficulty in calculating all parameters correctly in periods where the "load" becomes a generator such as with solar generation.

6. RESIDENTIAL TOU PRICING TRIAL

6.1 Introduction

The objective of this project was to determine the degree to which electric customers would alter their pattern of electricity consumption in response to electric prices that vary with the time of day, as well as measure their response to pricing that is triggered by predicted system critical peak conditions. To achieve these goals, this effort involved a 20-month experiment applying two distinct TOU pricing programs. Running from March 2013 through October 2014, this experiment initially included 62 homes in a Pricing Group and another 62 homes in a Control Group.

Most of the participants in this study reside in a neighborhood known as the Mueller development which is situated about three miles from downtown Austin. The neighborhood is built on space vacated by relocation of the old municipal airport. Construction of single-family homes began in August 2007. The homes have energy efficient features, and the neighborhood has a high concentration of installed grid-tied solar photovoltaics and plug-in electric vehicles (PEVs). Electric and natural gas distribution is provided to the neighborhood. Basic demographic information and typical features of the homes are listed in the table below:

Table 16. Demographic Data of Participants in Pricing Group and Control Group

Demographic Information	Responses
Salary	<ul style="list-style-type: none"> 43 survey respondents, or about 81%, claimed income above \$75,000. One respondent claimed income between \$300,000 and \$1 million, and two participants claimed income above \$1 million.
Ethnicity	<ul style="list-style-type: none"> 51 respondents, or 82%, reported Caucasian ethnicity. Three respondents reported Hispanic ethnicity. Three reported Asian/Pacific Islander ethnicity. Five declined to respond.
Education	<ul style="list-style-type: none"> All respondents, except for one left blank, reported a college degree or higher. 42 respondents, or 71%, reported a postgraduate degree.
Home during weekdays	<ul style="list-style-type: none"> 46% of the Pricing Group and 62% of the Control Group homes are typically occupied on weekdays (daytime hours).
PEV ownership	<ul style="list-style-type: none"> 34 plug in electric vehicles (PEVs) are owned by the Pricing Group, consisting of 25 plug-in hybrids and 9 full electric vehicles compared to only 6 PEVs for the Control Group. Note - 34 PEVs in the Pricing Group compared to only 6 in the Control Group is a large difference, affecting load profile and comparative results so the analysis had to account for this difference. Note - 34 PEVs in 62 homes is a much greater penetration than seen in the national market, providing a glimpse of potential future load shaping via shifting of vehicle charging schedules.

Demographic Information	Responses
Cooking appliances	<ul style="list-style-type: none"> Predominantly natural gas. 27% have an electric oven.
Laundry appliances	<ul style="list-style-type: none"> 1 in 3 has an electric dryer even though natural gas is available.
HVAC systems	<ul style="list-style-type: none"> 90% have a gas furnace.
Water heaters	<ul style="list-style-type: none"> Natural gas
Average square footage	<ul style="list-style-type: none"> Approximately 2,000 square feet for both groups.
% of homes with remotely programmable thermostats	<ul style="list-style-type: none"> 100% of homes in both groups.

The pricing experiment consisted of two pricing programs which did not coincide or overlap at any time:

- A seasonal TOU pricing experiment featuring a discounted night-time price per kilowatt-hour to encourage use of available inland (west Texas) wind generation during the five windiest months of the year (March through May, November and December), and
- A critical peak pricing (CPP) experiment invoking an expensive price per kilowatt-hour during designated peak hours on “critical peak” days during four summer months (June through September). Critical peak days were called a day ahead by the project Team based on forecasted next day weather conditions under the premise that extremely hot weather would stress the system, calling for voluntary load reductions.

The homes within the Pricing Group and the Control Group were outfitted with home energy monitoring (HEM) systems. These devices attached to the home’s individual electrical circuits within the breaker panel(s) and collected whole home and circuit (appliance) level energy use data in one-minute intervals. This data was made available to the consumers through a web-based interface so they could review the respective circuit-level energy use. This energy use data was used by the project to perform comparative analysis between and within the Pricing and Control Groups to characterize and quantify the Pricing Group’s response to the two experimental pricing programs.

Translation of the results of these experiments from this test group to the general population of Texas will require adjustments for many factors, including weather, demographics, home construction, residential distributed solar production, and market penetration of PEVs and home automation. The analysis described herein evaluates consumer behavior by “energy-savvy” residents with energy efficient homes, appliances, and remotely programmable thermostats. Therefore, it is intentionally more representative of a “future community” where such features are more common.

6.2 Anticipated Benefits

Broadly, the anticipated benefits of these pricing experiments were centered on gaining a better understanding of the effectiveness of pricing drivers to aid in management of and response to increased wind resources in Texas (TOU trial) and critical system peak events (CPP trial). Information gained in

these experiments was expected to support the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage in response to pricing drivers. The approach being tested here encourages customers to effectively and reliably manage the timing of their energy consumption as well as their peak demand in order to reduce their electricity bill. System benefits of positive customer response will be an increased coincidence of system load and night time wind generation, with corresponding reduced environmental impacts by offsetting generation using fossil fuels, and reduced system-wide capacity needs.

The two pricing experiments attempt to quantify the behavioral response of residential customers to pricing drivers. The seasonal TOU rate structure encourages shifting of consumption to the overnight period (10 p.m. and 6 a.m.) during certain months. The other is a voluntary demand reduction rate structure that goes into effect for a few days of the year (June – September) upon day-ahead notification when an ERCOT system critical peak condition is judged to be a strong possibility. These pricing trials give an indication of the degree to which the test group of customers can be expected to be aware and responsive to pricing drivers of the two types.

6.3 Interactions with Key Stakeholders

Over the course of the demonstration project, findings from the trials were presented and discussed at team meetings and public forums. Stakeholders include policy makers and interested parties, including ERCOT, Austin Energy, other utilities and retail electric providers, as well as members of the academic and energy-related products and services communities. Concurrent complimentary studies (e.g. residential PEV charging profiles) provide additional value to the body of research when taken as a whole.

Other key stakeholders include the actual pricing trial participants. In an interview with a sample of participants, they shared their approaches to shifting energy usage and ideas for adjusting the pricing regimes for improved effectiveness.

6.4 Technical Approach

This section provides a detailed description of the technical approach. The first part, Project Plan, identifies the basic pricing program rate structures, how customers were incentivized to participate, and the communications strategy employed to recruit participants. The second part, Data Collection and Benefits Analysis, defines the data collection and management, data visualization and analysis.

6.4.1 Project Plan

The following provides information on the overall project plan.

6.4.1.1 Incentive Account Tracking

Over the course of this 20-month experiment (March 2013 – October 2014), the Pricing Group and Control Group received and paid their usual Austin Energy electric bills, which reflected the normal Austin Energy electric rates.

A monetary incentive was used to entice the Pricing Group participants to either shift their energy consumption to the “wind enhancement” time period (10 p.m. – 6 a.m. during certain months) or to reduce energy consumption during CPP times (4 p.m. – 7 p.m. in the summer months). The project established a credit account with an initial balance of \$200 for each participant in the Pricing Group. The credit account balance at any point in time, whether above or below that initial \$200 amount, reflected the participant’s degree of response to price signals. Participants were able to view their credit account through an online web portal display as shown in the example figure below.

PERIOD	USAGE	PRICING TRIAL BILL	AUSTIN ENERGY BILL	ADJUSTMENT CREDIT
Mar 1, 2013 - Mar 31, 2013	674 kWh	\$59	\$62	\$3
Apr 1, 2013 - Apr 30, 2013	703 kWh	\$58	\$67	\$9
May 1, 2013 - May 31, 2013	658 kWh	\$54	\$58	\$3
Jun 1, 2013 - Jun 30, 2013	1050 kWh	\$128	\$138	\$10
Jul 1, 2013 - Jul 31, 2013	1088 kWh	\$114	\$128	\$14
Total	4173 kWh	\$413	\$453	\$239.0

Figure 58. Example of Web Portal Display of Representative Pricing Information for One Participant

The value of the credit in the account was adjusted every month at the time the participant received his or her electric bill. The “adjustment credit” reflected the difference between what the participant actually paid Austin Energy and what they would have paid if Austin Energy actually charged the pricing trial rates. Account balances provided feedback to the participants regarding their results in this experiment.

6.4.1.2 Critical Peak Pricing Rate Structure and Trial Design

The rate structures for both TOU programs were designed to be revenue neutral. This meant that, on average, if a participant chose to make no changes in their usage behavior, they would see no savings but they also would not be penalized. The participants were not really informed of this approach, it was just intended to ensure that the real focus was on measuring behavioral changes.

For the CPP trial, a price of \$0.64 per kilowatt hour (kWh) was applied to the Pricing Group participants' energy consumption during the weekday (Monday – Friday) afternoon hours of 4 p.m. – 7 p.m. on “critical peak” days during the four summer months, June through September. To make the experimental pricing revenue neutral, there was an offsetting discount of approximately \$0.016 per kWh on energy consumption during all non-peak hours of CPP event days. In 2013, 12 CPP days were called, and in 2014, 15 CPP days were called.

The plan was to call a maximum number of CPP events each month, according to the tables below, based on a minimum threshold temperature. The number of CPP events was adhered to during the project, but for 2014, the thresholds were revised downward due to it being a cooler summer.

Table 17. 2013 Goals and Thresholds for Calling Peak Days

Month	Threshold Temperature (°F)	Maximum Monthly Peak Events
June	102	3
July	103	4
August	103	5
September	102	3

Table 18. 2014 Goals and Thresholds for Calling Peak Days

Month	Threshold Temperature (°F)	Maximum Monthly Peak Events
June	101	3
July	101	4
August	103	5
September	100	3

The day-ahead CPP call was made by Frontier Associates to simulate a call by a utility or a retail electric provider anticipating an extremely high next-day system load. The following diagram presents the process that was employed to determine how and when to call a CPP event and notify the participants.

CPP Decision Flow Diagram:

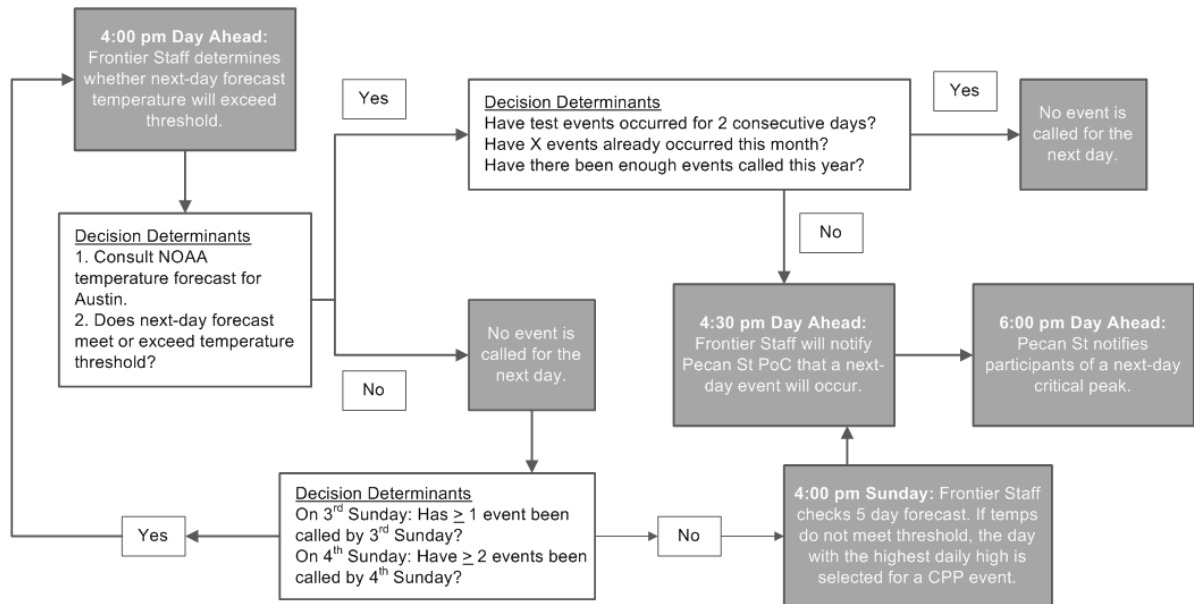


Figure 59. Flowchart for Calling Next-Day CPP Events

By notification of a next day CPP event, participants were able to take appropriate measures to shift energy use to outside the peak time frame the following day to avoid the high electricity rates during those three hours.

6.4.1.3 Wind Enhancement Rate Structure and Trial Design

For the wind enhancement pricing trial, the night-time electric rate for the Pricing Group during the five windiest months (March, April, May, November, and December) was lowered to \$0.0265 per kWh. This price applied to energy consumption in the hours from 10 p.m. – 6 a.m. on all days during those months. To make the experimental pricing revenue neutral, there was an offsetting surcharge of approximately \$0.02 per kWh on the participants' energy consumption during all other hours of those five months.

6.4.1.4 Calculation of Pricing Trial Electric Bills

The pricing trial electric rates consisted of these components:

1. Customer charge (the same as Austin Energy's charge) – a flat rate of \$10.
2. Baseline (existing Austin Energy) rates – applied only to consumption during non-wind hours and non-CPP hours.
3. Wind period rates – applied to consumption during wind hours
4. Surcharge rates – applied to total monthly consumption during non-wind hours
5. Discount rates – applied to consumption on CPP event days during hours other than 4 p.m.

through 7 p.m.

The pricing trial rate formulas were:

1. Wind Enhancement Period:

Customer charge + [Baseline rates x (consumption between 6 a.m. – 10 p.m.)] + [Wind Price x (consumption between 10 p.m. - 6 a.m.)] + [Surcharge Rate x (consumption between 6 a.m. – 10 p.m.)]

2. CPP Period:

Customer Charge + [Baseline rates x (consumption during all hours other than during called critical peak days, 4 p.m.-7 p.m.)] + [Critical Peak Price x (consumption during 4 p.m.-7 p.m. on called critical peak days)] + [Discounted rate x (consumption during all hours on event days other than 4 p.m. – 7 p.m.)]

If a Pricing Group participant's bill differed from his or her actual Austin Energy bill, the difference, whether positive or negative, was added to his or her credit balance. Once again, the participants started with a \$200 credit adjustment balance. At the end of the pricing trial, participants with positive credit adjustment balances were entitled to cash payments equal to the amount of the final credit adjustment balance in their accounts, up to a maximum of \$700 per customer. Participants with negative credit adjustment balances did not owe anything.

6.4.1.5 Participant Communication Strategy

Recruitment of pricing trial participants began in February 2013 with an outreach program designed to target residents living at the Mueller community in Austin. Owners of single-family residential homes were contacted via e-mail and asked for their participation in the TOU pricing trial. Most residents contacted were part of an earlier Pecan Street Energy Internet Demonstration (EID) program funded by another DOE cooperative agreement. These participants had the requisite HEM systems and legal participation agreements in place. The primary goal of recruiting from the EID pool for the pricing trial was to get a representative sample of participants that had a PEV.

Before the official start of the experiment in March 2013, the Pricing Group received e-mail communications outlining the experiment design and ideas for shifting or reducing their energy consumption. For instance, ideas included shifting laundry use to after 10 p.m. during the five wind enhancement months or reducing air conditioner use during CPP hours during the four summer months. The Team held one in-person workshop for the Pricing Group to further detail the trial's design, goals, and research benefits. Each member of the Pricing Group was given secure access to a personal online web portal to check the monthly status of their energy usage, their pricing trial bill, Austin Energy bill, and their adjustment credit balance.

On the web portal, in addition to pricing trial information, the Pricing Group participants had access to the following information:

- Monthly whole home energy use in kWh
- Monthly energy cost in U.S. dollars per appliance
- Energy generation cost in U.S. dollars if the participant has solar panels
- Real-time energy consumption in kWh
- Monthly energy cost comparison to other participants within the same zip code
- Monthly energy usage trends

During the summer months, Pricing Group participants also received cell-phone Short Message Service (SMS or “text”) messages and e-mail communications by 6:00 p.m. the day before a CPP day. The sample text message content shown below was designed to be strictly a factual reminder about the upcoming CPP day and the pricing rate.

Message: Tomorrow is a Critical Peak Pricing event
Your experimental electric rate will be
\$0.64 per kilowatt hour from 4 p.m. - 7 p.m.
Pecan Street Inc.

Figure 60. Example TOU Group Text Message

6.4.2 Data Collection and Benefits Analysis

6.4.2.1 Energy Data Collection

The HEM system installed in each home was a commercially available, off-the-shelf product extensively tested against other competing systems installed in 100 single-family homes during a one-year test bed trial in 2011 (as part of another project). The results of the tests found the selected HEM system to have superior reliability, durability, accuracy and customer support. Licensed master electricians installed a majority of the HEM systems under the prior EID program and maintained the HEM devices for the Pricing Group participants. Participating homes had one or two HEM systems installed, based on their unique circuit configuration.

Each HEM system is composed of:

- HEM device (energy data monitor)
- HomePlug adapter (power line communication device) allowing communications from the HEM system to the participant’s Internet gateway
- Up to twenty-four (24) 50-amp current transformers (CTs) for measuring consumption on individual circuits within the home

The HEM system directly measures electricity use for the whole home and a number of circuits at one-minute intervals. Individual circuits typically monitored include:

- Refrigerator
- Furnace and air handler
- Clothes dryer
- Clothes washer
- Oven and cook-top range
- Microwave
- Electric vehicle level II charger
- Lighting - whole home
- Lighting - individual room
- Plug loads
- Solar generation (if installed)

The HEM system allowed the collection of one-minute interval disaggregated data at a resolution that could detect granular changes in the energy use of individual appliances which was critical in characterizing behavioral changes in response to pricing signals. The data from the HEM system was streamed through the participants' Internet gateway to Pecan Street's server cluster where it was stored in a secure database environment.

6.4.2.2 Data Management and Cyber-Security Considerations

As of October 2014, the database contained over 183 million records and continues to grow at a rate of 3 million records per day. Given the size and diversity of data, Pecan Street established a highly available, scalable database architecture that adheres to the organization's DOE-approved cyber-security plan protecting the identity of all participants while organizing the data so that it is accessible to appropriate parties. A data warehousing strategy was employed that integrates the energy use, participant and survey data sets into common schema or systems, with emphasis on commonalities in the data such as participant, measurement type, and time-stamp.

Due to the intimate and confidential nature of some data being collected, data security is a critical aspect of the data integration and storage. As such, the main security principle applied is protection of the project participants' identities and privacy. The database adheres to a Cyber-Security Plan approved by DOE which protects the identity of all research participants and organizes the data so that it is accessible to approved researchers. The plan requires that any personally identifiable information must be kept in a separate database from all other data on the home and linked by a master database accessible only to Pecan Street staff via a 2-factor authentication process.

6.4.2.3 Data Visualization

The Team created a series of anonymized data visualizations that are rendered through an online web portal to communicate the energy data collected from Pricing and Control Group participants. These visualizations show energy load profiles for the groups and a geospatial “heat map” rendering of energy consumption that is scaled and generalized at the community level in order to protect against identification of individual home locations.

6.4.2.4 Data Analysis

The analysis methods are described below, and the results are presented in section 6.5. Conclusions are summarized in section 6.6.

6.4.2.4.1 CPP Trial

Data on electricity usage profiles on critical peak days was analyzed using three different statistical methods, specifically looking for evidence of reduction in usage between 4 p.m. and 7 p.m. on called event days. The first method discussed below, “fixed panel data analysis,” is considered the primary method, yielding the most reliable results¹¹. The other two methods, also discussed below, “historical baseline method” and “difference in difference method,” were used for comparison purposes only.

1. Fixed effects panel data analysis:

The Team performed panel data analysis, taking time and weather into consideration. This method segregates the weather effects from the behavior change during the peak event, giving more representative results. Panel data are a set of variables measured on multiple observational units at multiple time periods. In this case, a whole summer dataset is analyzed including both the Pricing Group and the Control Group. Fixed effects panel data analysis, rather than random effects panel data analysis, is applied when time independent effects for each entity are correlated with the predictors. In this case, only parameters that vary over time are taken into consideration, such as hourly temperature and whether or not it is a peak period. The model was initially designed as follows:

$$(kW)_{it} = \alpha_i + \beta_1*(cdh)_t + \beta_2*(Event)_{it} + \beta_3*(hourstart17)_t + \beta_4*(hourstart18)_t + \sum_{k=5}^{10} \beta_k*(dayofweek)_t + \varepsilon_i$$

Where:

- cdh” indicates the cooling degree hours from 4-5 p.m., 5-6 p.m., and 6-7 p.m. (cooling degree hours = max(hourly temperature-70,0))

¹¹ This method is also employed more frequently in analysis of dynamic rate pilots (re: Orlando, Ernest, Lawrence Berkeley National Laboratory, “Quantifying the Impacts of Time-based Rates, Enabling Technology, and Other Treatments in Consumer Behavior Studies: Protocols and Guidelines,” LBNL-6301E , July 2013)

- “i” is each individual (each Pricing Group and Control Group member)
- “t” is each summer afternoon hour (16:00-17:00, 17:00-18:00 and 18:00-19:00 each day from June 1st to September 30th)

For 2013, based on 12 events, it was implemented as:

$$(kW)_{it} = \alpha_i + \sum_{k=1}^{12} \beta_k * (cppevent_k)_{it} + \beta_{13} * (cdh)_t + \beta_{14} * (Event)_{it} + \beta_{15} * (hourstart17)_t + \beta_{16} * (hourstart18)_t + \sum_{k=17}^{22} \beta_k * (dayofweek)_t + \varepsilon_{it}$$

For 2014, based on 15 events, it was implemented as:

$$(kW)_{it} = \alpha_i + \sum_{k=1}^{15} \beta_k * (cppevent_k)_{it} + \beta_{16} * (cdh)_t + \beta_{17} * (Event)_{it} + \beta_{18} * (hourstart17)_t + \beta_{19} * (hourstart18)_t + \sum_{k=20}^{25} \beta_k * (dayofweek)_t + \varepsilon_{it}$$

2. Historical baseline analysis:

A historical baseline analysis, along with the difference in difference (DID) analysis, was used to provide alternative savings estimates for reference and comparison to panel data analysis results. For each CPP event, estimated kW savings were based on the following equation:

kW savings = Baseline Average Peak kW – CPP Period Average Peak kW

Where:

- Baseline Average Peak kW = average kW demand recorded 4:00 p.m. - 7:00 p.m. during the four non-peak weekdays immediately preceding the called CPP day
- CPP Period Average Peak kW = average kW demand recorded 4:00 p.m. - 7:00 p.m. on the called CPP day

The example in the table below shows the kW savings calculation for a single event day - 20 June 2013. Note that in this example, the fourth preceding weekday actually occurred the prior week and since it was much cooler, it was not included in the average baseline calculation.

Table 19. Example of Critical Peak Pricing Day kW Savings Calculation (Day Matching Method)

Day	Temperature	Baseline Period kW	CPP kW	kW savings
20-Jun	98		2.0999587	
19-Jun	98	2.897029		
18-Jun	95	2.350909		
17-Jun	99	2.884796		
14-Jun*	93	2.085456		
Average		2.710911	2.0999587	0.610953

* Not used in average (unseasonably cool day)

3. Difference in difference analysis:

Difference in difference (DID) analysis, along with historical baseline analysis, were used to provide alternative savings estimates for reference and comparison to panel data analysis results. DID analysis can be used when hourly load data for all customers in the treatment group and control group are available before and during the treatment period. We can use pooled cross-sectional data to isolate and quantify the effects of these rate treatments on one or more measures of load response.¹² A simple linear regression can be used to perform the analysis:

$$Y = \alpha_0 + \delta_0 D_2 + \alpha_1 D_m + \delta_1 D_2 D_m + \varepsilon$$

Where:

- Y = hourly kW for each household
- D_2 = a dummy variable that takes on a value of unity if the observation is from the treatment period and zero otherwise
- D_m = a dummy variable that takes on a value of unity if the observation is for a member of the treatment group and zero otherwise

The coefficient of interest is δ_1 because it defines the interaction between the time period and the treatment dummies, $D_2 * D_m$.¹³ In other words, coefficient δ_1 estimate is the kW savings estimate we were looking for.

Event day 4:00-7:00 p.m. is used as the treatment period and the previous non-event weekday (4:00-7:00 p.m.) is used as the control period.

6.4.2.4.2 Wind Enhancement Pricing Trial

Data was analyzed primarily by calculating and comparing the average percentage and kWh daily consumption per home during the period of 10 p.m. to 6 a.m. to that during the period of 6 a.m. to 10 p.m. The 24 hour usage distribution was examined to quantify the degree of shifting accomplished by the participants in response to the seasonal TOU pricing which encouraged night time consumption.

Shifting of specific loads was evaluated in similar fashion for expected candidate end uses, such as vehicle charging, kitchen appliance use, and laundry-related tasks.

6.5 Performance Results

6.5.1 Operation of Smart Grid Technologies and Systems

Detailed analysis results are reported in this section. They are summarized in bullet form in section 6.6, "Conclusions."

¹² Orlando, Ernest, Lawrence Berkeley National Laboratory, "Quantifying the Impacts of Time-based Rates, Enabling Technology, and Other Treatments in Consumer Behavior Studies: Protocols and Guidelines," LBNL-6301E, July 2013

¹³ Ibid.

6.5.1.1 Overall Results - Adjustment Credit Account Analysis

Each Pricing Group participant's credit adjustment of \$200 was adjusted monthly based on their respective TOU pricing results. The analysis results described below depict the adjusted credit balances (not including the initial \$200 credit).

Overall, there was a positive response to pricing signals as indicated by the levels of adjustment credit earned by the Pricing Group participants. 97% of participants had positive credits while 3% (two participants) had deductions on their incentive account. 84% of participants had earnings greater than \$40 (in addition to the \$200 incentive credit) and the maximum adjustment credit earned was \$280.57, not including the initial credit. The average adjustment credit earned was \$125.13, for an average payout of \$325.13 to participants in this experiment.

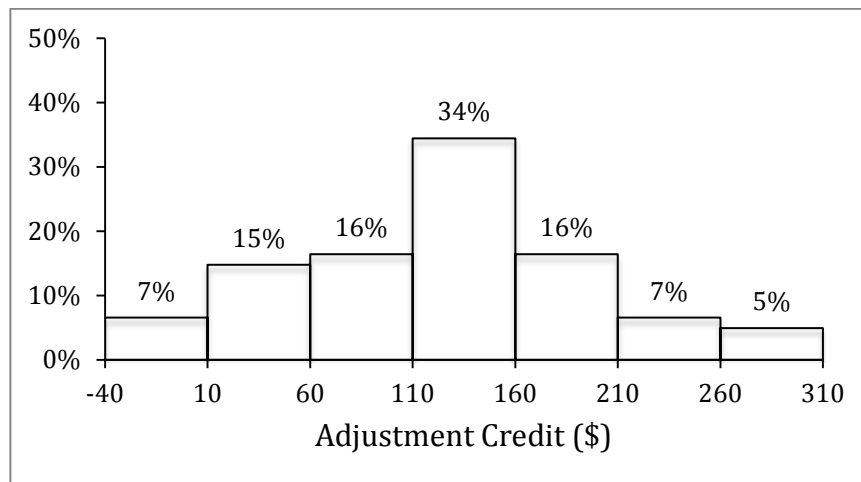


Figure 61. Adjustment Credit Earnings Distribution

The graphs in the figures above and below reflect the distribution of incentive credits earned by the participants over the course of the pricing trial. On the lower end, approximately 7% of the Pricing Group participants (4 out of 61) earned less than \$10 in adjustment credit (in addition to the \$200 initial credit). On the higher end, roughly 5% of the Pricing Group participants (3 out of 61) earned greater than \$260 in credit. The predominant amount of credit earned over the course of the project was in the range of \$110 to \$160 (not including the \$200 incentive credit). Even though the largest incentive amount earned was significantly below the maximum allowable amount of \$700, overall there was a positive response to the TOU pricing programs.

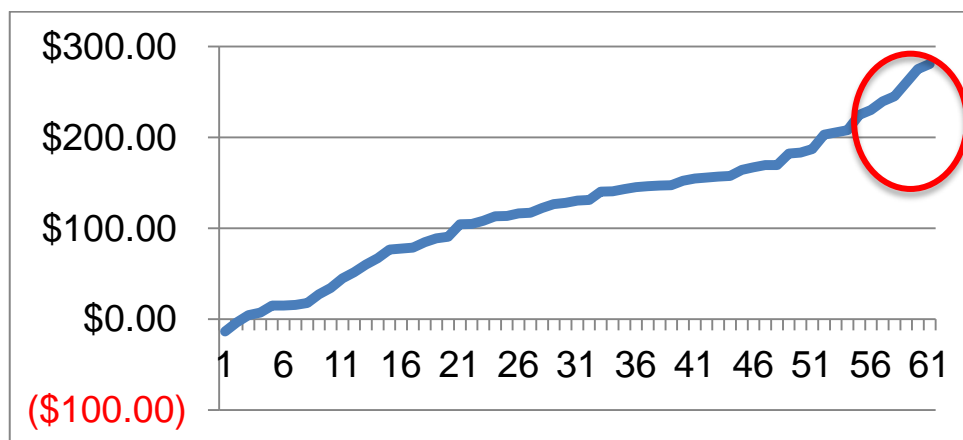


Figure 62. TOU Incentives Payout Distribution

For the five participants who received the largest payouts, the Team compared their savings during the wind months versus the summer months. The results are shown below. These participants generally earned the majority of their credit in the wind pricing months, except for one (circled).

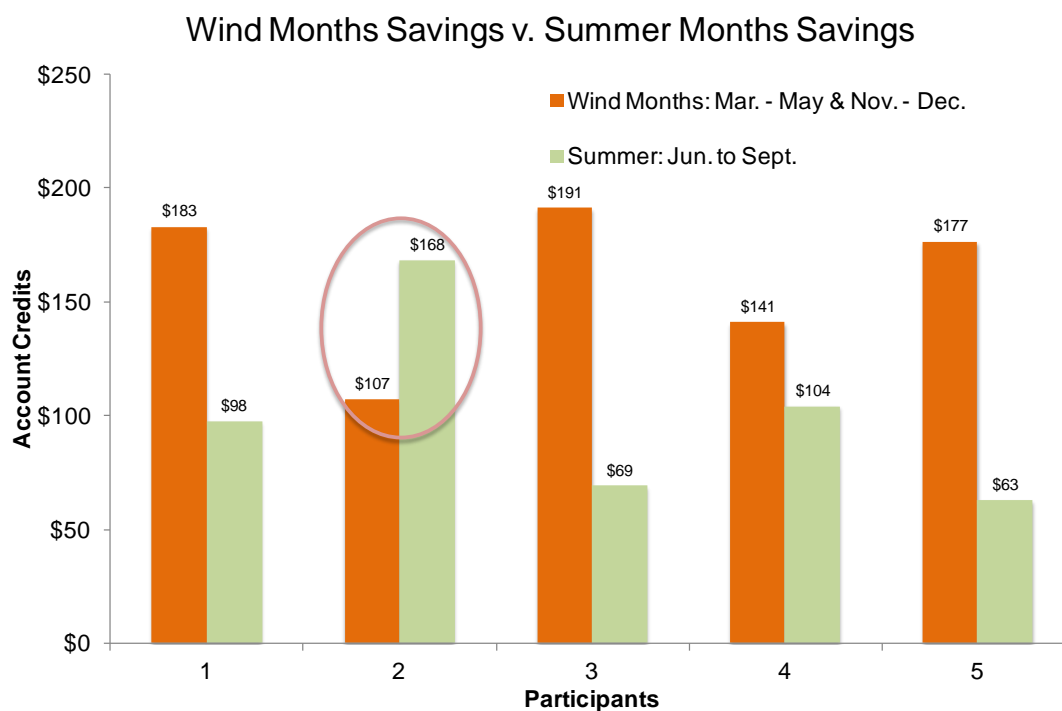


Figure 63. Wind Months Savings versus Summer Months Savings

This analysis was repeated for those participants who received the 10 median payouts. The comparison results are shown below. These participants also generally earned the majority of their credit in the wind pricing months, except for one (circled).

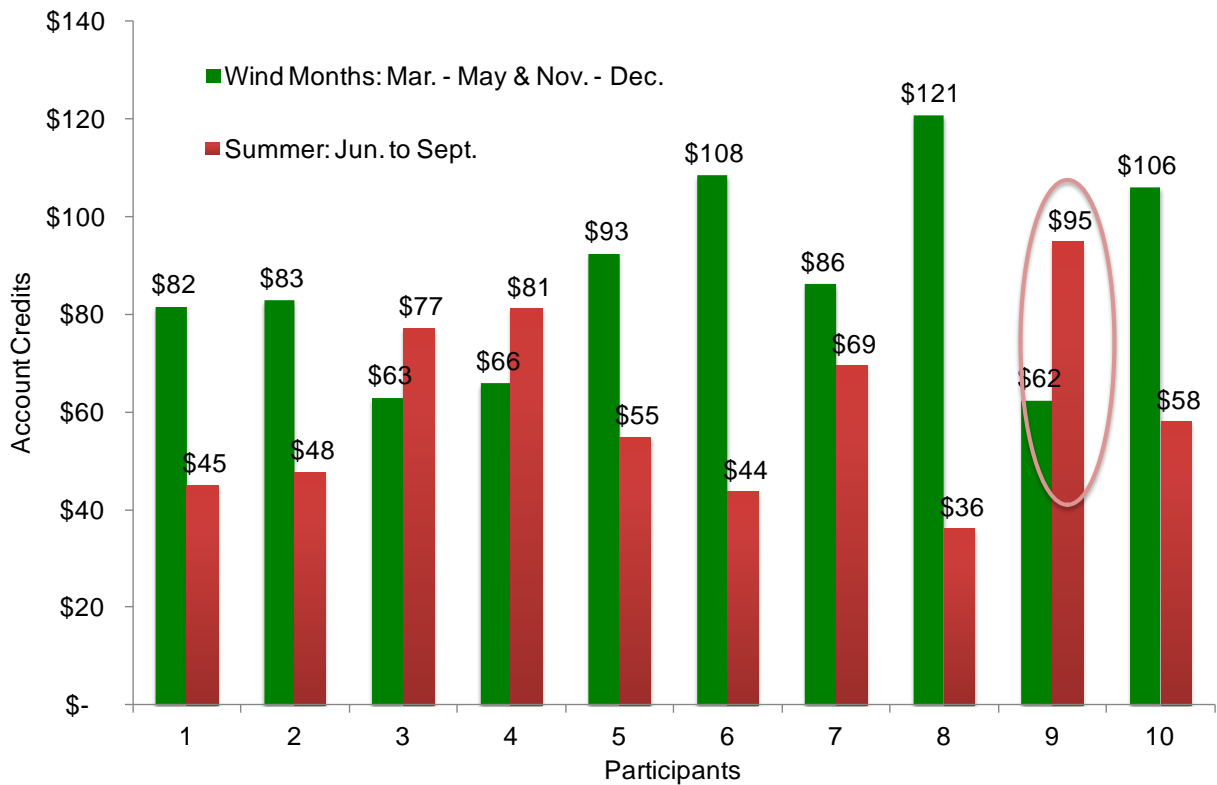


Figure 64. Median Payout Accounts – Wind Months Savings versus Summer Months Savings

6.5.1.2 CPP Trial Results

Results are presented below for the three statistical methods used to estimate the demand reduction for critical peak days. The first method presented, “fixed panel data analysis,” is considered the most reliable and thus the primary method of analysis. Specific observations are reported on the breakdown of end uses contributing to CPP reductions, PEV owner CPP response, and a comparison of CPP responses for homes that are occupied versus unoccupied during weekday afternoons.

6.5.1.2.1 Panel Data Analysis

Using panel data analysis as the primary analysis method applied to hourly consumption data, Frontier found the following CPP demand reductions per event:

2013: 0.44 kW or 15.27% reduction in demand

2014: 0.17 kW or 6.15% reduction in demand

The tables below provide further data on each CPP event day.

Table 20. 2013 Panel Data Analysis Results

Critical Peak Day	Daily High (F)	Estimated kW Savings
June 20th	98	0.5475
June 26th	101	0.7455
June 28th	104	0.7002
July 24th	97	0.4194
July 26th	99	0.3590
August 1st	103	0.2100
August 7th	103	0.7047
August 8th	104	0.6429
August 29th	99	0.2768
August 30th	102	0.2897
September 5th	96	0.1259
September 13th	97	0.3272
Average Savings per Event:		0.44 kW

Table 21. 2014 Panel Data Analysis Results

Critical Peak Day	Daily High (F)	Estimated kW Savings
June 24th	93	-0.2484
June 30th	96	0.2840
July 2nd	97	0.0068
July 14th	97	0.2064
July 24th	101	0.0052
July 28th	99	0.1574
August 7th	102	0.2292
August 15th	101	0.1420
August 19th	96	0.5093
August 21st	103	0.5163
August 26th	101	0.2264
September 3rd	100	0.1601
September 9th	97	0.2082
September 11th	99	-0.0185
September 15th	88	0.2618
Average Savings per Event:		0.17 kW

As one can see from the tabulations above, CPP events in 2014 achieved less peak reductions on average than in 2013. The figure below graphically illustrates the range and variability of results for the two years.

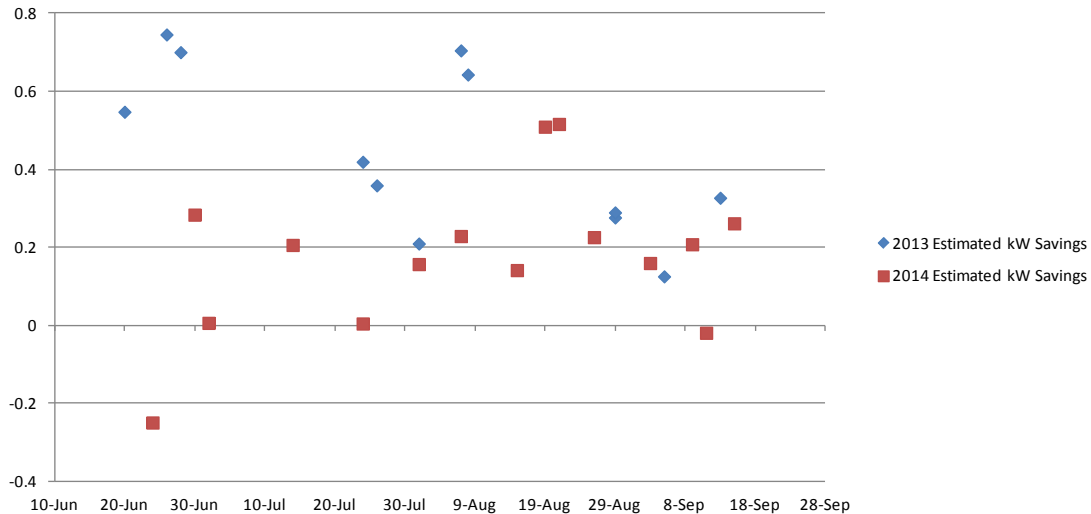


Figure 65. 2013 versus 2014 CPP Events Savings Comparison

Although the absolute magnitude of the peak reductions averaged less than 0.5 kW, the percentage peak reduction of 15% in 2013 is in line with other studies. Regarding the reductions averaging less than 0.5 kW, it is important to note that according to final survey responses from the Pricing Group, participants made lifestyle changes early in the trials that shifted load away from the peak hours. The changes were persistent throughout the entire trial period, thus diminishing the CPP reductions since the baseline (non-CPP summer day) loads were already reduced by load shifting. Examples include shifting of PEV charging, laundry and kitchen appliance use to after 10 p.m. when possible, as a matter of routine.

The differing levels of response in the summer of 2013 versus 2014 bear further discussion. Responses to the final survey of the Pricing Group indicated that many participants experienced “customer fatigue” during the second summer (2014). Respondents reported competing priorities and disappointing electricity cost savings. One quarter stated that they did not participate in the second summer’s CPP trial at all, and it is plausible that the others participated to a lesser degree the second summer for similar reasons.

In consideration of these performance factors:

- The 2013 results were used in breakdown analysis.
- The CPP peak reductions reported here can be considered incremental to the on-peak load shifts realized in a TOU pricing environment, whereby participants already have a somewhat reduced baseline load during peak hours.
- It can be expected that voluntary CPP programs like this one that rely on deliberate participant action will produce widely varying responses and will struggle to maintain persistence.

The next section reports results of two alternate analysis methods for quantifying CPP kW reductions: “historical baseline method” and “difference in difference method.” They provide diverse approaches to the calculations.

6.5.1.2.2 Results from Alternate Analysis Methods

For comparison purposes only, the historical baseline method and difference in difference method were used as alternate methods for calculating the CPP kW savings. Results for the multiple analysis methods are tabulated below.

Table 22. CPP Peak Reduction Results

	Estimated 2013 Peak Reduction	Estimated 2014 Peak Reduction
Historical baseline method	0.29 kW	0.085 kW
Panel data analysis	0.44 kW	0.17 kW
Difference in difference method	0.498 kW	0.245 kW

As seen above, the alternate estimates envelope the values from the panel data analysis, which is considered the most reliable method for estimating kW demand savings from this study.

6.5.1.2.3 Breakdown Analysis of CPP kW Savings

Further investigation was conducted to determine which end-uses contributed to the peak load reduction during the 12 CPP events in 2013. Peak load reduction was postulated as most likely coming from the following end-use appliances, for which participants could voluntarily reduce usage in the 4 p.m. to 7 p.m. window: HVAC, PEV charging, kitchen appliances and laundry-related usage.

Among these end-uses, the first two categories were found to contribute the most to the energy reduction. Considering the fact that 40% of the participants in the Pricing Group have PEVs, the Team conducted this part of the analysis by first dividing the group into two subgroups: with and without PEVs. Since HVAC usage is weather-sensitive, sub-metered HVAC consumption was used with panel data analysis to estimate average HVAC kW reductions for each CPP event. For those who own PEVs, the day-matching (“historical baseline”) method described in prior sections was used to calculate PEV kW savings for each CPP event. For the subgroup owning PEVs, kW savings breakdown analysis results are tabulated and graphed on the following pages.

Table 23. 2013 Breakdown Analysis of kW Savings - Participants with PEVs

	Estimated Total kW Savings	HVAC kW Savings	Estimated PEV kW Charging Savings	The Rest
June 20th	0.5390	0.1714	0.1123	0.2553
June 26th	0.8075	0.3481	0.3234	0.1360
June 28th	0.8816	0.4032	0.2225	0.2559
July 24th	0.3298	0.1325	0.0446	0.1527

July 26th	0.4097	0.1774	0.0928	0.1395
August 1st	0.2023	0.1253	-0.2143	0.2913
August 7th	0.6167	0.252	0.1462	0.2185
August 8th	0.8449	0.3014	0.2019	0.3416
August 29th	0.1939	0.0633	0.1057	0.0249
August 30th	0.1377	0.0064	0.2312	-0.0999
September 5th	0.2317	0.0923	0.0100	0.1294
September 13th	0.3102	0.1358	-0.0465	0.2209
Average	0.459	0.184	0.102	0.173
% of Total	N/A	40.13%	22.34%	37.53%

Table 24. 2013 Breakdown Analysis of kW Savings - Participants without PEVs

	Estimated Total kW Savings	HVAC kW Savings	The Rest
June 20th	0.4181	0.2499	0.1682
June 26th	0.5674	0.3725	0.1949
June 28th	0.458	0.2570	0.2010
July 24th	0.3504	0.2262	0.1242
July 26th	0.1574	0.0864	0.0710
August 1st	0.1506	0.1315	0.0191
August 7th	0.3623	0.2702	0.0921
August 8th	0.2627	0.0991	0.1636
August 29th	0.2259	0.0343	0.1916
August 30th	0.1276	0.0384	0.0892
September 5th	0.0618	0.1368	-0.0750
September 13th	0.2465	0.0445	0.2020
Average	0.282	0.162	0.120
% of Total	N/A	57.45%	42.55%

From the two tables above, for Pricing Group members with PEVs, around 62% of the CPP kW reductions were attributed to HVAC and PEV related behavior change (HVAC @ 40%, PEVs @ 22%). For Pricing Group members without PEVs, around 57% of the CPP kW reduction was attributed to HVAC related behavior change.

Apart from HVAC and PEV charging reductions, the other 40% or so of total CPP kW reductions came from other end uses. The Team conducted in-depth analysis on sub-metered laundry and kitchen appliance consumption and found both to be significant contributors, with additional reductions distributed unremarkably throughout multiple end uses.

6.5.1.2.4 PEV Owners Reduce Peak Load More than Non-PEV Owners

In general, Pricing Group members with PEVs saved more energy during CPP events (0.46 kW versus 0.28 kW) than non-PEV owners, as depicted in the graph below for the 2013 events.

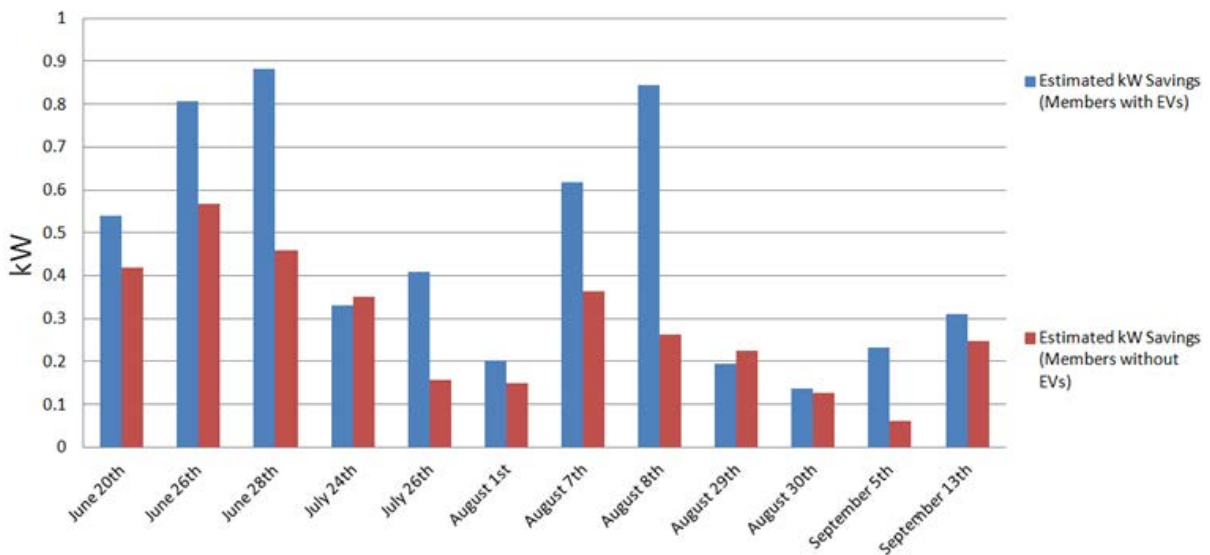


Figure 66. Overall 2013 kW Savings Comparison of CPP Participants With and Without EVs

The Pricing Group participants who own PEVs also exhibited more variability in responding to CPP events. For those who charged in the afternoons, if they delayed that charging until after 7:00 p.m. during CPP events, their peak load reductions were much higher than for those without PEVs. PEV charging is a high kW load that can often be delayed until later in response to pricing. See section 6.5.1.3.2 for a comprehensive discussion of PEV charging response.

6.5.1.2.5 Occupied versus Unoccupied Homes Energy Usage During CPP Events

The graph in the figure below shows that homes that are typically occupied on weekday afternoons in both the Control Group and Pricing Group used more energy overall than homes that are normally unoccupied at those times. The Pricing Group homes managed significant reductions during CPP events, with the occupied homes turning in the most impressive CPP performance.

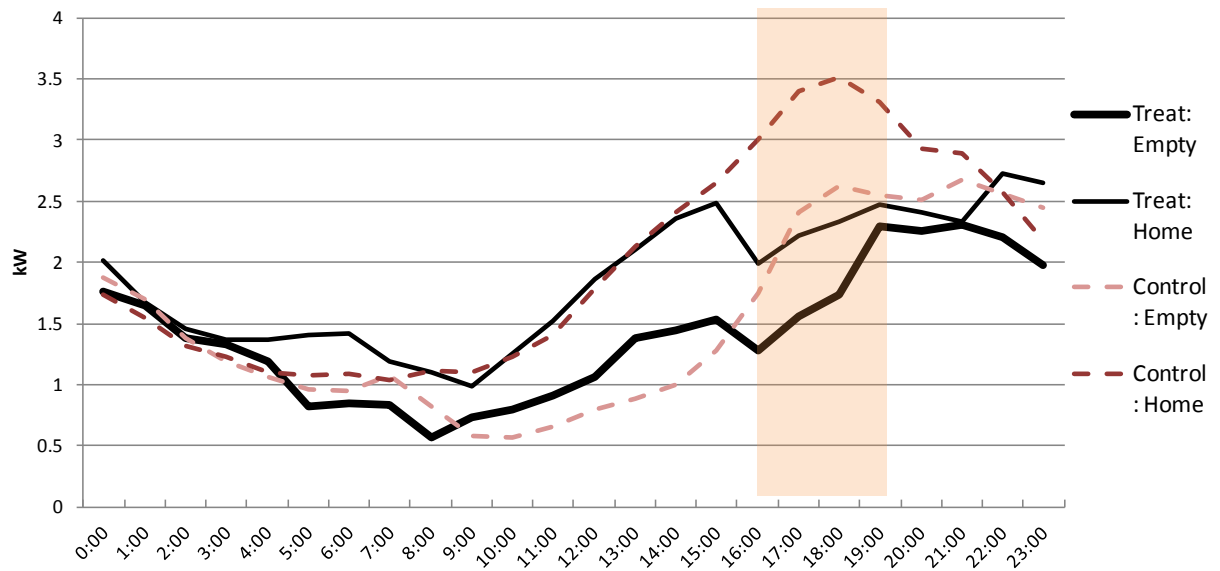


Figure 67. 2013 CPP Savings for Occupied versus Unoccupied Homes

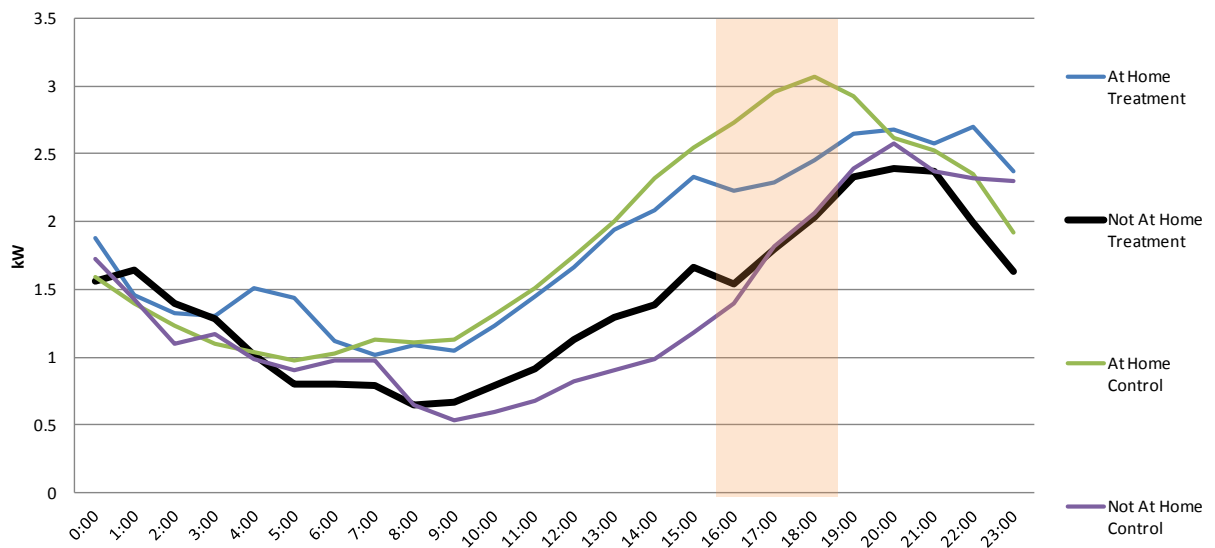


Figure 68. 2014 CPP Savings for Occupied versus Unoccupied Homes

The graphs above allow a meaningful comparison of occupied (“Home”) and unoccupied (“Empty”) homes for the Control group and the Pricing (“Treatment”) Group during the peak time frame of 4 p.m. to 7 p.m. on CPP event days. Note that the occupied/unoccupied load shapes for the Pricing Group and the Control Group are virtually identical until the 4 p.m. hour.

The Control Group has no response in the 4 p.m. to 7 p.m. time frame as they are not informed of the CPP event, thus indicating default behavior in the absence of CPP program participation. It can be observed that the occupied homes naturally use more energy during the day than unoccupied homes, likely indicating that the use of programmable thermostats is reducing energy cost for unoccupied homes.

In contrast to the default behavior of the Control Group, note the significant response of the Pricing Group at the 4 p.m. hour for both the occupied and unoccupied homes. Both exhibit marked step reductions at 4 p.m., indicating that (1) residents who are home take action to consume less energy, primarily by reducing HVAC and PEV charging usage as reported elsewhere in this report, and (2) residents who are not home also take action on those days to reduce HVAC usage, likely once again using their programmable thermostats, in some cases accessing them remotely to change settings, as indicated in participant feedback.

Note that the magnitude of the response to a critical peak event call is slightly greater for homes where the residents are home. This is because these homes have both higher default consumption and the means and opportunity to reduce consumption by manual action.

6.5.1.2.6 No Conservation Effect from CPP Trial

No reduction in total energy consumption was found on called critical peak days versus non-event days.

6.5.1.2.7 Comparison with Other CPP Trials

Frontier Associates conducted a literature review of journal articles about CPP trials across North America.¹⁴ The results were as follows:

¹⁴ Faruqui, Ahmad and Sanem Sergici. “Household response to dynamic pricing of electricity: a survey of 15 experiments.” The Brattle Group. 31 August 2010. Published Online.

“Understanding Electric Utility Customers – Summary Report.” 2012 Technical Report by EPRI.

Faruqui, Ahmad and Jenny Palmer. “Dynamic Pricing of Electricity and its Discontents.” The Brattle Group. August 3, 2011.

Wolak, F. A. (2006). Residential customer response to real-time pricing: The Anaheim Critical-Peak

Faruqui, Ahmad and Sanem Sergici. “Arcturus: International Evidence on Dynamic Pricing.”

Faruqui, Ahmad, Sanem Sergici, and Neil Lessem. “Impact Evaluation of Ontario’s Time-of-Use Rates: First Year Analysis.” Prepared for the Ontario Power Authority. The Brattle Group. November 26, 2013.

Faruqui, Ahmad and Sanem Sergici. “The Power of Experimentation: New evidence on residential demand response.” The Brattle Group. May 11, 2008.

- Compared with other CPP trials, the demand reduction measured in this trial is on the low side. CPP tariffs typically induce a drop in peak demand between 13 and 20%. Customers with central air conditioners display a response in the 15 to 20% range. For customers with enabling technologies, such as remotely accessible and programmable thermostats, which were generally available to the participants in this trial, the reduction grows to between 25 and 45 percent.
- Some pilots found persistence over multiple years. One pilot experiment in Maryland found that customers actually became more price-responsive in the second summer of the trial. However, most articles made no conclusions about the persistence of the effects of the pricing scheme over several years.

6.5.1.2.8 CPP Trial - Final Survey of Participants

In November 2014, after the study period ended, additional input was sought from Pricing Group participants via a final survey with a few targeted questions. 55% responded and the responses relating to the CCP trial are summarized here:

- Most participants made lifestyle changes the first few months of the trial in response to TOU wind pricing, shifting load away from afternoon peak hours to after 10 p.m. where feasible, e.g. PEV charging, operating laundry, and running the dishwasher. Many of these changes persisted throughout the entire trial period. Note: This persistent shift decreased the measured CPP reductions in summer months.
- 80% of the Pricing Group participated in the 2013 (first summer) CPP program. Of the active participants:
 - One quarter did not participate in the second summer's CPP trial
 - Many respondents reported "fatigue" from participating in the program. Two respondents mentioned the difficulty of maintaining participation with small children.
 - Virtually all reduced their HVAC use, almost half of them by pre-cooling the home before the event period started (4 p.m.) on called event days.
 - About half adopted some of their CPP response actions into their summertime routine *every day*, not just on event days.
 - 20% of respondents who were at home during a typical weekday afternoon reported that they intentionally left the house during a called event period. Note that this result is surprisingly high and would not likely persist long term.

6.5.1.3 Wind Enhancement Pricing Trial Results

Seasonal TOU pricing, specifically night time "Wind Enhancement Pricing" in this study, was put in place for the months of March to May and November to December in 2013 and 2014 (the wind pricing months) in an attempt to cause participants to shift some of their usage of electricity to the overnight hours to be served by west Texas wind generation. The Team compared the Pricing Group's 24-hour electricity consumption pattern during these months to the Control Group's consumption pattern.

Consumption patterns for certain sub-metered end uses (PEV charging, kitchen appliances and laundry) were also examined to characterize any observed shifts.

Due to the differing levels of PEV ownership between the Pricing and Control Groups – 34 PEVs (56%) in the Pricing Group versus only 6 PEVs (10%) in the Control Group – the Pricing Group’s consumption patterns were also compared to their own pre-trial consumption patterns.

6.5.1.3.1 Response to TOU Pricing - Overall Consumption

The figures below compare Pricing Group to Control Group daily consumption patterns during the 2013 and 2014 wind pricing months, plotted in hourly percentage of daily total usage. The seasonal average hourly Texas wind generation is represented by the dotted line in the background.¹⁵ Note that shifting electricity consumption from day time to overnight supports better matching of load to wind resources.

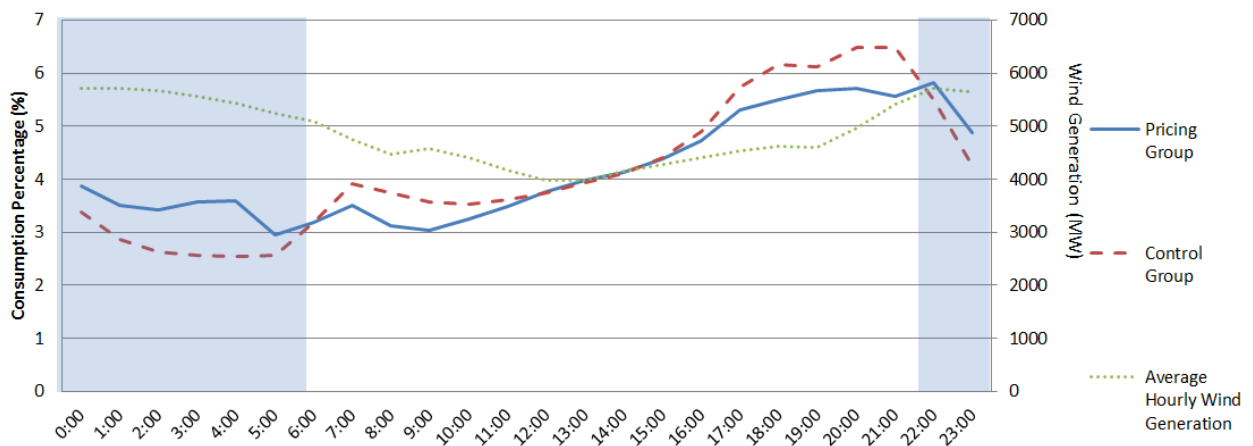


Figure 69. Pricing Group versus Control Group 2013 March - May Wind Pricing Period

¹⁵ Wind Generation Data Source:

<http://mis.ercot.com/misapp/GetReports.do?reportTypeId=13424&reportTitle=Hourly%20Aggregated%20Wind%20Output&showHTMLView=&mimicKey>

Note: The curves shown include inland and coastal wind generation

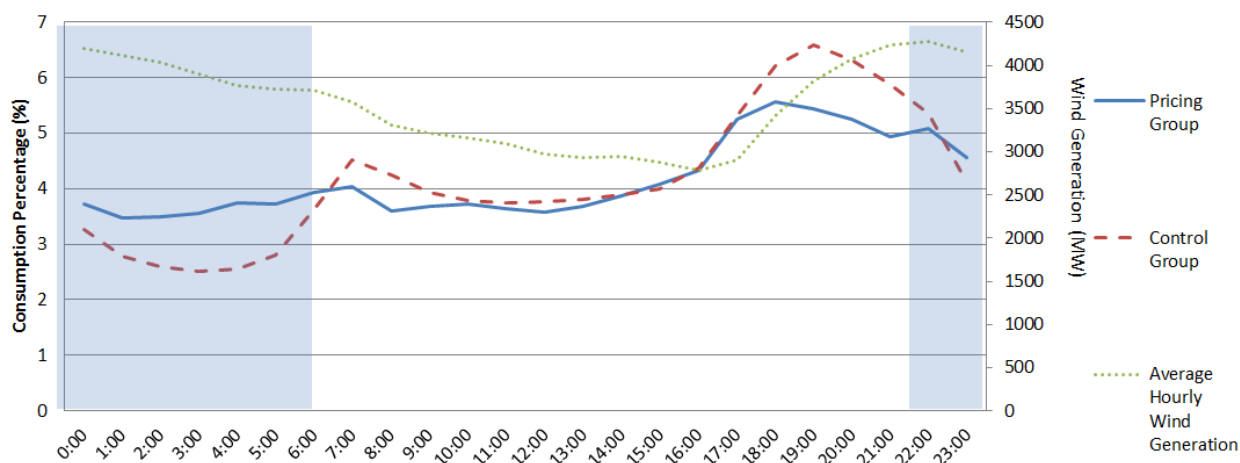


Figure 70. Pricing Group versus Control Group 2013 Nov-Dec Wind Pricing Period

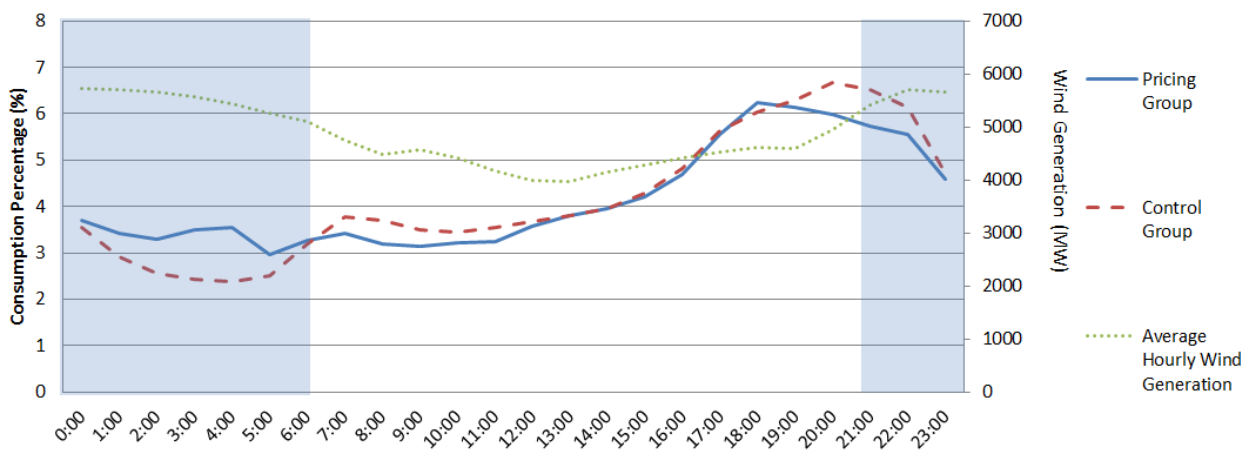


Figure 71. Pricing Group versus Control Group 2014 March – May Wind Pricing Period

In contrast to the shifts evident in wind-pricing months (above), there is a weaker shift in non-wind months (see graph below for June 2014).

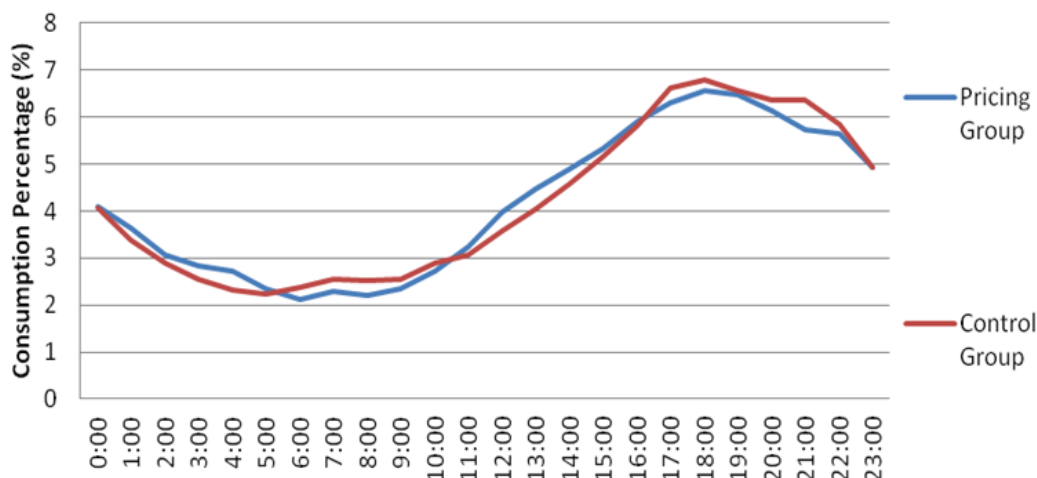
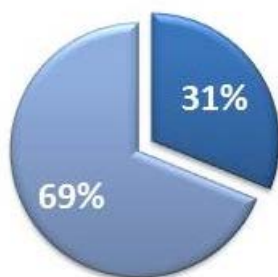


Figure 72. Pricing Group versus Control Group June 2014

Daytime/night time splits of overall electricity consumption for the Pricing and Control groups are compared below, indicating a 4% difference, which is small but statistically significant.

Pricing Group All Wind Periods

10:00pm-6:00am 6:00am-10:00pm



Control Group All Wind Periods

10:00pm-6:00am 6:00am-10:00pm



Figure 73. Pricing Group versus Control Group Consumption - All Wind Pricing Months

The 4% finding is determined to be 4-5% based on discussions summarized in the next section.

6.5.1.3.2 Response to TOU Pricing – PEV Charging

This 4% difference in day/night consumption between the Pricing and Control Groups (27% versus 31% as seen above) cannot directly be assumed to be a load “shift” because of the differing levels of PEV ownership between the groups (56% or 34 PEVs in the Pricing Group, only 10% or 6 PEVs in the Control Group). Type 2 PEV charging, at 240 volts and approximately 6 kW, or approximately equal to the demand of a 5-ton air conditioner, is the largest single consumption that can be shifted to night time. Furthermore, due to the widespread pre-trial practice of charging their PEVs throughout the day, as

found for participants in these trials, it is meaningful to compare the Pricing Group's pre-trial consumption to its own trial period consumption (figure below).

Pricing Group Pre-Trial Usage (2012 Nov-Dec)

■ 10:00pm-6:00am ■ 6:00am-10:00pm



Pricing Group Overall Trial Usage (2013 Nov-Dec)

■ 10:00pm-6:00am ■ 6:00am-10:00pm



Figure 74. Pricing Group versus Self, Pre-Trial versus Trial Period

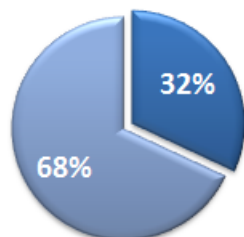
The 5% shift of consumption to night time (26% to 31% shown above) due to the TOU pricing is slightly greater than the 4% difference between the Pricing and Control Groups. This led us to conclude that there is a 4-5% shift to overnight use by the Pricing Group.

The percentage of PEV ownership in the Pricing Group affects the amount of day time load available to shift to night time. The following discussion examines PEV charging behavior, as affected by TOU pricing.

The percentage of PEV charging done at night saw a very impressive shift to night time in response to the TOU pricing for both plug in hybrids and full electric vehicles, as depicted by the pie charts below. This indicates that influencing the timing of PEV charging is important for shaping load.

Pricing Group PEV Charging (Pre-Trial)

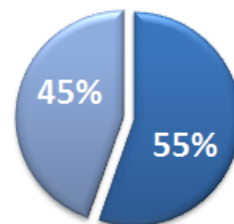
■ 10:00pm-6:00am ■ 6:00am-10:00pm



Nov.2012-Feb.2013

Pricing Group PEV Charging (Trial)

■ 10:00pm-6:00am ■ 6:00am-10:00pm



Mar-May & Nov-Dec 2013 and Mar-May 2014

Figure 75. Pricing Group PEV Owners Shift in Percent Night Time Charging

The average PEV charging consumption daily load shape is depicted in the next two graphs for owners of plug-in hybrids and full electric PEVs, for three different periods—before the trial started, during all wind months, and during the summer. It was suspected that charging behavior would differ between the two types of PEVs, so they are graphed separately.

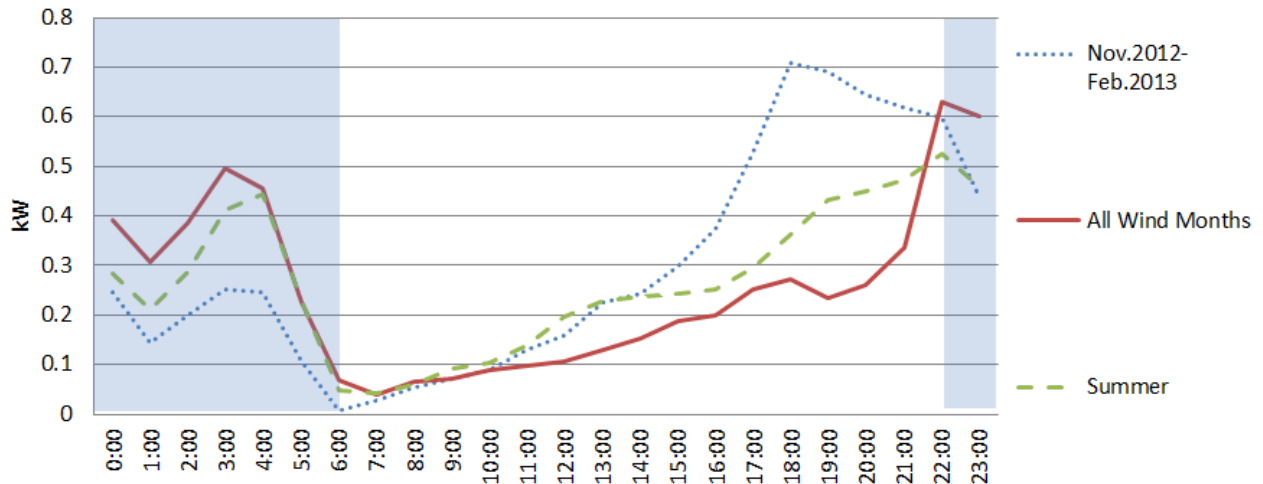


Figure 76. Pricing Group Hybrid PEV Owners Average Charging Behavior

In the figure above, (dotted line versus solid line) note the marked reduction in day time charging by hybrid owners when the TOU wind pricing is in effect, taking advantage of very inexpensive overnight charging. Also note the partial rebound in day time charging during the summer months (dashed line) when wind pricing is not in effect. This summer rebound may be partially caused by the longer daylight hours and increase in evening activities in the summer time, calling for increased seasonal afternoon charging due to the short range of hybrids. Some participants stated in the final survey responses that they went back to charging when it was convenient (including daytime hours) when the TOU pricing was not in effect.

A similar graph is shown below for the full electric PEV owners.

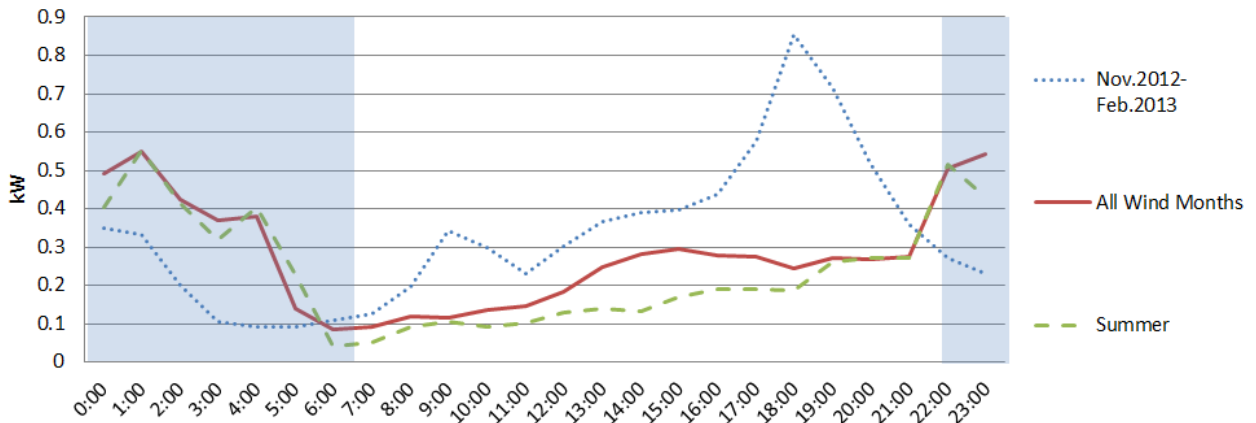


Figure 77. Pricing Group Full Electric PEV Owners Average Charging Behavior

In the above figure, the full electric PEV owners also showed a marked reduction in day time charging when the TOU wind pricing is in effect (dotted line versus solid line).. But note that the summer rebound is not seen in this case (dashed line), likely because full electric PEVs often have the range needed to handle the evening tasks, allowing charging to be scheduled for overnight.

Because summer on-peak charging is of interest for system planning, and since charging is not highly correlated to weather, it is useful to compare the charging behavior of plug-in hybrid and full electric PEV owners in the 4 p.m. to 7 p.m. window, with and without TOU wind pricing.

Table 25. PEV Charging Response to Pricing

Period of Interest	Average kW Weekday non-Holiday 4 p.m. – 7 p.m. Plug in Hybrid PEV Owners	Average kW Weekday non-Holiday 4 p.m. – 7 p.m. Full Electric PEV Owners
Pre-Trial (default behavior)	0.5376 kW	0.6226 kW
TOU Wind Pricing Months	0.2414 kW	0.2655 kW
Summer (no TOU pricing)	0.3032 kW	0.1888 kW

It is likely that the average pre-trial charging load (0.6226 kW) for full electric PEVs was boosted by the higher typical charge rate for these vehicles versus some of the hybrids.

As noted in the preceding graphs and table, the pre-trial default behavior for both hybrid and full electric PEV owners was to charge for convenience during the day time hours, peaking at around 6 p.m. as reported elsewhere, upon arriving back home from work.¹⁶

Response to the TOU pricing was very pronounced for PEV owners as a group, resulting in greater than a 50% reduction in daytime charging and 4 p.m. to 7 p.m. charging. Night time charging nearly doubled.

Feedback from participant PEV owners indicated that the pricing incentives and educational outreach offered with these trials prompted them to learn how to program their vehicle for overnight charging, rather than charging at will during the day or simply manually starting to charge at 10 p.m. Not surprisingly, changes in charging behavior were found in this study to be persistent over the period of this study for most participants as they became part of the daily routine.

Relevant supporting research on PEVs revealed the following, giving more impetus to the need to influence the timing of PEV charging:

- 2013 total PEV Sales in US were 97,500 (1.6% of the 6 million total automobiles sold)¹⁷
- PEV sales are projected to grow at 16% per year between 2014 and 2023;¹⁸ Texas is expected to be a leading state for growth in PEV sales
- Sales of PEVs are running 50/50 between plug-in hybrid and full electric¹⁹ (breakdown in our participant group was 3:1 in favor of hybrids)
- With battery capacities of only 8-16 kWh, a fully charged plug-in hybrid uses all-electric drive for only the first 15-30 miles, then switches to series-parallel electric/combustion engine drive after batteries reach a minimum charge state. This short range encourages a pattern of frequent charging throughout the day.
- Longer range electric vehicles expected in the next 2-5 years,²⁰ with their higher charge rates (up to 15 kW) making it critical to understand and influence charging behavior.

6.5.1.3.3 Response to TOU Pricing – Other End Uses

In addition to PEV charging, the Team also analyzed the Pricing Group's pattern of daily consumption for individually sub-metered end uses, looking for loads being shifted to after 10 p.m. Kitchen appliance

¹⁶ Faruqui, Ahmad, Ryan Hledik, Armando Levy, and Alan Madian. "Will Smart Prices Induce Smart Charging of Electric Vehicles?" The Brattle Group. July 2011 (page 7)

¹⁷ Inside EVs, September 2014 Plug-In Electric Vehicle Sales Report Card, <http://insideevs.com/september-2014-plug-electric-vehicle-sales-report-card/> (accessed December, 2014)

¹⁸ Navigant Research, Gren Car Congress, December 1, 2014, <http://www.greencarcongress.com/forecasts/> (accessed December, 2014)

¹⁹ Inside EVs, September 2014 Plug-In Electric Vehicle Sales Report Card, <http://insideevs.com/september-2014-plug-electric-vehicle-sales-report-card/> (accessed December, 2014)

²⁰ Brad Berman, "The Coming Era of Long-Range Electric Cars," Plug-in Cars, Oct 19, 2014, <http://www.plugin-cars.com/coming-era-long-range-electric-cars-130203.html> (accessed 10/30/14) "300+ mi ranges by 2017" per Volkswagen official

use, water heating and laundry-related consumption would normally be expected to provide significant load shifting impact. In fact, participant surveys reported that dishwashing and laundry activities were often shifted to after 10 p.m. However, these changes did not produce dramatic shifts, largely because the Mueller community is served by natural gas, so the majority of cooking and water heating is fueled by natural gas.

Daytime HVAC use is not a candidate for shifting to overnight; also the wind months are shoulder months (low heating and cooling energy required). So, aside from shifting of PEV charging, water heating and laundry related loads to night time, the balance of load shifting was observed to be split unremarkably among various end uses of electricity in these homes.

6.5.1.3.4 No Conservation Effect from Wind Enhancement TOU Trial

The Team calculated the total annual electricity use of the Pricing Group on a rolling 12-month basis, looking for a downward trend which would indicate a “conservation effect” developing during the trial. None was noted.

Applying a different perspective, the Pricing Group’s average kWh consumption in August 2013 (1,387 kWh) was compared to August 2014 (1,379 kWh). The two are virtually identical without weather adjustment, despite August 2014 being a milder month, again indicating no conservation effect.

6.5.1.3.5 Comparison to Other TOU Pricing Trials

This study estimates a 4-5% shift from day time to overnight consumption as reported above. This response is in alignment with the 3-6% impacts found in other studies.²¹

6.5.1.3.6 Wind Enhancement Trial - Final Survey of participants

In November 2014, after the study period ended, additional input was sought from Pricing Group participants via a final survey with a few targeted questions. 55% responded, and the responses relating to the TOU wind pricing trial are summarized here:

- Just over half of survey respondents indicated they altered their energy use to take advantage of cheaper overnight power during the TOU trial.
 - It is interesting that only HALF of a volunteer trial group actively participated by trying to shift load to save money.
 - Most active participants made lifestyle changes at the beginning of the trial in response to TOU wind pricing, shifting load away from afternoon peak hours, to after 10 p.m. where feasible, e.g. charging PEV, doing laundry, and running the dishwasher.
 - 80% of those who actively participated said that the trial resulted in persistent changes to their energy consumption patterns. This indicates they practiced this “learned”

²¹ Faruqui, Ahmad and Sanem Sergici. “Household response to dynamic pricing of electricity: a survey of 15 experiments.” The Brattle Group. 31 August 2010. Published Online

behavior during the CPP trial months as well, thereby reducing the average kW reduction of the Pricing Group in the CPP trial.

- Virtually all PEV owners in these programs stated that they shifted their car charging to alternate times, but the measured impact of this shift is diminished by the fact that many participants had already shifted their car charging to after 10 p.m. as a matter of routine during the wind pricing trial.
- One third shifted their laundry and dishwasher activities to after 10 p.m. Participants would have realized greater savings had this community not featured - almost exclusively - natural gas furnaces, water heaters and clothes dryers.

6.5.2 Metrics and Benefits Analysis

6.5.2.1 Operational Benefits

Critical peak pricing (CPP) is a tool for motivating the customer, using avoidance of high dynamic electric rates, to curtail or shift load away from peak hours on days when high loads are predicted to approach or exceed system capacity. Customers voluntarily sign on to a special plan with a CPP structure, typically with the customer's goal being to save money by opportunistically adjusting their consumption in response to critical peak events. This is beneficial to the utility system and to the customer, helping to avoid system emergencies.

TOU wind pricing in this study offers to save the customer money when the customer shifts consumption to night time. This increased night time load helps utilize west Texas wind generation that peaks at night. This shift decreases the impact on other generation assets trying to follow the system load at night (fossil-fired plants reducing load or cycling on/off) and can reduce afternoon system peak loads.

6.5.2.2 Build Metrics and Benefits

The build metrics focus on hardware and software items expended to support the project. For this pricing effort, about \$449,312 was expended for hardware and software computing resources and licenses to support the testing and data analysis efforts, and \$11,250 was spent on the purchase of 50 NEST thermostats for in-home testing.

6.5.2.3 Impact Metrics and Benefits

Metrics for rigorously estimating the potential impact of the pricing incentives, as found in this study, on the Texas market would be based upon extrapolating and adjusting the predicted responses to homeowners with more typical residential building styles and demographics. Several prerequisites and adjustments would be needed:

- Prerequisites
 - Advanced metering (interval meters)

- Sufficient pricing incentive to induce behavior change
- Adjustments
 - Weather
 - HVAC equipment variables – mix of heating types (electric strip, heat pump, gas furnace), efficiency (heat pump), and use of programmable thermostats
 - Quality of construction and condition of the home
 - Conditioned square footage
 - Demographics – ability and likelihood of pricing response
 - Green commitment
 - Mix of electric versus natural gas or liquid propane fueled appliances (water heating, clothes drying, cooking)
 - PEV penetration and mix of vehicle types (plug in hybrid versus full electric)
 - Degree of acceptance of special rate plans of this nature

6.5.2.3.1 CPP Impact

- Critical Peak Pricing reduced peak consumption on event days by the following amounts:

Summer 2013: 0.44 kW or 15.27% reduction in demand

Summer 2014: 0.17 kW or 6.15% reduction in demand

- Because of interaction with the TOU pricing trial and second year “fatigue” reported by participants, a peak demand reduction of 0.44 kW (15%) per residence can be estimated from this CPP trial, with the qualifications that it is a reasonable impact estimate for the *initial year* of a CPP program, when the response to a *day-ahead CPP notification* is implemented in a “*non-automated*” fashion (without remote web- or AMI-enabled notification and curtailment action, i.e. fully relying on the customer to take action), in a demographic group such as represented in this study, in *energy efficient housing stock* with programmable thermostats, and in a pricing environment that *also features other incentives to move load off the afternoon peak*, such as TOU pricing. This CPP study’s ratio of peak to off-peak electricity rate was approximately 5:1.
- If a 15% peak reduction was realized for all residential load in ERCOT, which comprises about half of the estimated 68,000 MW system peak load for 2014, the peak reduction would be 5,100 MW. This study did not attempt to estimate *achievable* performance.
- Environmental impact – Since there was no significant reduction in total consumption on critical peak days versus other summer days, there is no environmental impact from CPP

6.5.2.3.2 TOU Wind Enhancement Pricing Impact

- It is meaningful to compare the megawatt hours (MWh) potentially shifted in response to seasonal “wind enhancement” TOU pricing to the amount of wind generation in ERCOT during the shoulder months.
 - Calculations show that if the TOU wind enhancement pricing trial were to be implemented statewide at every residential unit, successfully shifting about 5% of the 24-hour consumption during shoulder months to night-time periods, the shift would equate to roughly one third to one half of the seasonally available overnight wind generation currently available.²² This study did not attempt to estimate *achievable* performance.
 - In the future, it is anticipated that this percentage would trend lower, assuming wind generation continues to grow at a faster rate than residential loads.
- Environmental impact – With simple cycle combustion turbines providing marginal generation on peak, shifting of electricity consumption from daytime to night time may have estimable environmental impact. However, the modeling required for such estimations was beyond the scope of this study.

6.5.3 Stakeholder Feedback

6.5.3.1 Stakeholders

Stakeholders included policy makers and interested parties, including ERCOT, Austin Energy, other utilities and REPs, as well as members of the academic and energy-related products and services communities. These stakeholders have indicated that the findings of these trials will be useful in planning across Texas, informing continued study and decision making on the use of pricing programs such as CPP and TOU rates, and for addressing the challenges of increased market penetration of PEVs. The participants also provided feedback through their behavioral changes and also during surveys and interviews, and their opinions on these pricing programs and how they were executed will continue to inform other stakeholders.

²² Estimate uses 2012 EIA data on total Texas residential electricity consumption and ERCOT 2012 data on total wind generation. The proportion of residential electricity consumption occurring during shoulder months was estimated using ERCOT’s 2012 report of Actual Energy Use by Month (available at http://www.ercot.com/news/press_releases/show/26382, accessed 12/2/2014). The total amount of wind generation occurring during shoulder months was derived from ERCOT’s Energy by Fuel Type, 2002-2013 report (available at <http://www.ercot.com/news/presentations/>, accessed 12/2/2014). The proportion of wind generation occurring during nighttime hours of shoulder months was estimated using monthly wind generation profiles for west Texas wind in Assessment of Wind/Solar Co-Located Generation in Texas (available at <http://www.cleanenergyassociates.com/wp-content/uploads/2013/03/AE-Colocated-Solar-Wind-Study-20090923.pdf>, accessed 12/2/2014).

6.5.3.2 Participants

The following is applicable to both the CPP and wind enhancement pricing trials.

On December 10, 2013, a subset of residents participating in the pricing trial met with Pecan Street and Frontier to discuss the project and provide feedback. Participants were asked questions in an informal setting to gain more understanding of their responses to the pricing options and the text message alerts.

- Each of the residents present at the meeting had made changes to their electricity use at the onset of the project due to gaining more knowledge about the program itself and not necessarily due to the pricing triggers. Energy use that could be switched to overnight or off-peak hours was generally moved early in the trial period, if convenient, with electric vehicle charging being the biggest contributor to this shift. When convenient, other electric loads were also moved, such as dishwasher use or laundry activities.
- The residents generally reacted to the critical peak alerts with peak load reductions if possible and convenient. Access to a remote-controlled thermostat, which all the Pricing Group participants had, was key to reacting to these alerts by being able to change the set point temperature in their homes. If a participant worked from home, pre-cooling was often done and the thermostat was set higher during the 4 p.m. to 7 p.m. window. Otherwise, comfort was chosen over conservation.
- One resident who worked from home scheduled his daily energy use around the pricing trial and potential alerts, including pre-cooling the house before 4 p.m. every day, so that even if a critical peak day alert was called, no changes were needed.
- None of the interviewed residents stated that the program was too difficult to manage, or a nuisance. Again, convenience, comfort and conservation commitment were the deciding factors on program participation. The monetary incentive was more of a bonus for participating in the trial. Interviewees even mentioned that the 64 cent/kWh CPP rate might need to be increased to become a more effective incentive.

In November 2014, after the study period ended, additional input was sought from Pricing Group participants via a final survey with a few targeted questions. 55% responded, and their responses were previously reported in sections 6.5.1.28 and 6.5.1.3.6..

6.6 Conclusions

6.6.1 Projections of Demonstration and Commercial Scale System Performance

6.6.1.1 CPP Trial

This trial defined critical peak consumption as kWh consumed between 4 p.m. and 7 p.m. on critical peak days called a day in advance. An experimental rate of \$0.64/kWh was applied to this peak consumption, whereas the normal incremental rate is \$0.11/kWh.

Conclusions:

- Critical Peak Pricing reduced peak consumption on event days by the following amounts:
 - Summer 2013: 0.44 kW or 15.27% reduction in demand
 - Summer 2014: 0.17 kW or 6.15% reduction in demand
- Because of interaction with the TOU pricing trial and second year “fatigue” reported by participants, a peak demand reduction of 0.44 kW (15%) per residence can be estimated from this CPP trial, with the qualifications that it is a reasonable impact estimate for the *initial year* of a CPP program, when the response to a *day-ahead CPP notification* is implemented in a “*non-automated*” fashion (without remote web- or AMI-enabled notification and curtailment action, i.e. fully relying on the customer to take action), in a demographic group such as represented in this study, in *energy efficient housing stock*, and in a pricing environment that also features other incentives to move load off the afternoon peak, such as TOU pricing. This CPP study’s ratio of peak to off-peak electricity rate was approximately 5:1.
- A CPP regime such as implemented in this study - featuring day-ahead event calling and fully voluntary event day action required of participants - has drawbacks:
 - Critical Peak days are not always recognized one day ahead. This drawback indicates the need for a backup “same day” call regime, but communications and response on a same day basis are inherently problematic.
 - Planners must realize that peak reductions will be diminished by other peak load shifting programs in effect.
 - To be effective, the pricing must be punitive enough to strongly induce voluntary curtailment or shifting; otherwise, customers will find it not worth the inconvenience or discomfort of responding.
 - For a CPP program as structured in this study, to be effective in reducing on-peak load on days when peak reduction proves critical, the following must occur:
 - The entity calling the CPP event must correctly predict the weather and system operating parameters 24 hours in advance.
 - The customer must be *able* and *have the means* and *the will* to take physical action to reduce load at the proper time
 - It can be expected that a non-automated CPP regime such as used in this trial will produce widely varying responses from one call to the next and will struggle to maintain persistence.
- Critical Peak Pricing can be effective in shifting load for participants that own PEVs, once those participants receive prompting and training on how to program their EV chargers. However, it is highly desirable that PEV owners avoid on-peak charging on *all days* whenever feasible. This desired persistent behavior is more effectively encouraged by other types of programs such as TOU pricing. Once non-essential on-peak charging is minimized, it will be unavailable as a load shift in response to a critical peak call.

- The largest individual contributors to CPP demand reduction *for PEV owners* were found to be HVAC and PEV charging, as shown in the 2013 data presented below. Other contributors were laundry and kitchen appliance use (e.g. dishwashers) which can be shifted to off-peak run times.

Table 26. (2013) CPP Peak Demand Reduction per Event (PEV Owners)

	Estimated Total kW Reduction	HVAC kW Reduction	Estimated PEV kW Charging Reduction	The Rest
Average	0.45875	0.184092	0.102483	0.172175
% of Total	N/A	40.13%	22.34%	37.53%

The largest individual contributor to CPP demand reduction *for non-PEV owners* was HVAC.

Table 27. (2013) CPP Peak Demand Reductions per Event (Non-PEV Owners)

	Estimated Total kW Reduction	HVAC kW Reduction	The Rest
Average	0.282392	0.162233	0.120158
% of Total	N/A	57.45%	42.55%

The end use constituents of kW reductions will have a significantly different makeup in different weather zones, with different demographics, different market penetrations of PEVs, urban versus rural areas, with/without the availability of natural gas or propane for space and water heating and clothes drying, and with/without other pricing programs in place that shift load away from peak hours.

- Homes that are typically occupied on weekday afternoons use more energy overall than homes that are normally unoccupied at those times. Occupied homes turned in the most impressive CPP demand reductions in this study because these homes have higher default consumption (occupied on weekdays) and the opportunity to reduce consumption by occupants taking action.

6.6.1.2 Wind Enhancement Pricing Trial

This trial attempted to induce load shifting from daytime to overnight hours. A very low experimental price of \$0.0265/kWh was applied to consumption between 10 p.m. and 6 a.m. in the months (March-May and November-December) when inland wind generation is most available, while a surcharge was applied to consumption during the daytime. The ratio of peak to off-peak rates ("rate ratio") was about 4:1.

Conclusions:

- “Wind enhancement” TOU pricing caused a shift of 4-5% of the average daily consumption from day time to overnight hours. This finding is in alignment with the 3-6% impacts found in other studies ²³.
- A consumption shift of 5% across Texas is roughly estimated to utilize one third to one half of the seasonally available overnight wind generation.
- Pricing responses in different locales would be affected by weather zone, demographics, market penetration and daily miles traveled by PEVs (e.g. urban versus rural areas), and the availability of natural gas. This voluntary study group of participants was not a typical demographic. They were highly educated and “energy savvy”. Half of them were likely to be at home during weekdays. Their homes had a high technology penetration in general and were energy efficient. Most of them used natural gas for water heating and clothes drying, reducing the potential for electric load shifting. One third of them owned electric vehicles.
- A TOU pricing regime is an effective and reliable pricing program for moving load off-peak, as customers are constantly encouraged to shift load. They are rewarded for persistent behavior change, and such change can become a part of their daily routine.
- Persistence of beneficial behaviors is aided by technology (e.g. programmable thermostats, appliance timers, PEV charge scheduling).
- Education, outreach and reinforcement are important enablers of persistent change in energy consumption patterns.
- Electric vehicle (PEV) charging, typically at 3.3 kW or 6 kW, is the largest singular end use load that can be shifted to night time. But in the absence of pricing incentives, education and outreach, new PEV owners do not necessarily set a programmed charging schedule, let alone set charging for overnight hours.
- Default behavior of the PEV owners in this study, as found in other studies, was to charge the vehicle for convenience during the day, year-round, charging heavily during peak hours of 4 p.m. to 7 p.m., i.e. upon returning home from work. This study found that daytime charging is just as prevalent among full electric PEV owners as among hybrid PEV owners.
- With pricing incentives and prompting, PEV owners will shift most of their daily electric vehicle charging to overnight. If the pricing incentives are seasonal, some of the PEV owners, especially hybrid owners, will revert to daytime charging when the pricing incentives are not in effect.
- PEV charging behavior change results are broken down in the table below for hybrid and full electric PEVs. The reductions from “Pre-Trial” to “Wind Pricing” months indicate response to pricing. The difference between “Pre-Trial” and “Summer” months indicates persistent positive learned behavior, aided by technology (charge scheduling), executed even in the absence of TOU pricing.

23 Faruqui, Ahmad and Sanem Sergici. “Household response to dynamic pricing of electricity: a survey of 15 experiments.” The Brattle Group. 31 August 2010. Published Online

Table 28. PEV Charging Demand Changes

Period of Interest	Average Charging kW Weekday non-Holiday 4 p.m. – 7 p.m. Plug-In Hybrid Owners	Average Charging kW Weekday non-Holiday 4 p.m. – 7 p.m. Full Electric PEV Owners
Pre-Trial (default behavior)	0.5376 kW	0.6226 kW
TOU Wind Pricing Months	0.2414 kW	0.2655 kW
Summer (no TOU pricing) ²⁴	0.3032 kW	0.1888 kW

- Regional PEV market trends and owner charging profiles should be studied and monitored for system planning and infrastructure impacts, particularly as longer range vehicles come to market in the next 2-5 years.
- In summary, maximizing residential PEV charging overnight will:
 - Support increased use of available Texas inland wind generation
 - Require compelling pricing incentives and education/outreach
 - Be sensitive to PEV market penetration and type of vehicles
 - Mitigate potential system peak impacts (on-peak charging)

6.6.2 Lessons Learned and Best Practices

The following outlines the major lessons learned and best practices from these pricing trials.

- Planners must account for the interactive effects of multiple pricing programs running concurrently or sequentially. One or both programs can produce lesser measured results than would be expected if implemented in isolation. This is demonstrated by the results from these CPP and TOU pricing trials, which were conducted with the same participants over alternating periods. The apparent on-peak kW reductions from the CPP trial were significantly reduced by persistent behavior changes first elicited by the preceding TOU trial. The two pricing programs were chasing some of the same candidate end use load shifts – electric vehicle charging, kitchen appliance usage (cooking, dishwashing) and laundry-related usage (washer, dryer, water heating).
- As the TOU trial was a more consistent savings opportunity available for months at a time, whereas the CPP trial involved only a few days over the entire summer, the participants saved more from the TOU trial. They responded to the TOU trial by putting in place the scheduling and behavior change needed to capitalize on inexpensive

²⁴ As noted previously, in the conclusions from the CPP trial, PEV owners FURTHER reduce their charging by another 0.102 kW on called critical peak days in the summer.

overnight power when available. These changes, or learned behavior, remained in place throughout the study period for many participants, as indicated in the final survey responses, thus diminishing the measured load reductions from the CPP trial.

- Not all customers who sign up for a pricing program will actually participate, and changes in their situations can cause them to drop out over time. Participants can suffer from ‘customer fatigue’. Response to pricing programs appears to be enthusiastic and successful in the beginning when the program is new and the educational and outreach messages are fresh. As mentioned by participants in the final survey responses, interest can fade over time, other priorities take over, perhaps savings are not perceived to be high enough for the inconvenience involved, and the incentives begin to have less direct effect on behavior. These factors point to a need for intense and continuous outreach, feedback and reinforcement.
- PEV charging is vital for shaping load patterns. Electric vehicles are the largest load in a residential home that can be scheduled with flexibility. Most charging can be performed overnight, when the vehicle is not in use, and the charge will often be sufficient for the activities of the following day. Educating PEV owners about charge scheduling is a big first step in shifting these loads. Without such instruction, charging times are erratic and concentrated in the afternoon-evening hours (4 p.m. – 10 p.m.). Participants indicated that they do not immediately learn how to program the charging times when they purchase the vehicle. In these trials, for homes with a PEV, charging load accounted for nearly 25% of the peak load shift on critical peak days and was a key element in shifting load to overnight hours during wind pricing months.
- CPP programs that call events “real-time” with demand response actions that are automated (with opt out provision) and do not call for physical action by the participant²⁵ should produce more impact and be more persistent than a program such as the one implemented in this trial, which requires a deliberate response by the homeowner.
- Programmable HVAC thermostats, and in fact all appliances with programmable scheduling, are powerful tools for shifting load. They can be set up to take advantage of TOU rates like the wind-enhancement and critical peak rates in this trial.
- Reliable, interoperable communications are essential. Uninterrupted data communications are needed to ensure reliable meter data collection and system feedback. Data caching is needed to compensate for downed wireless connections.

²⁵ e.g. Gulf Power Energy Select pricing program; http://smartgridcc.org/wp-content/uploads/2013/07/CS_Gulf-Power.pdf (accessed December 2014)

7. RESIDENTIAL DEMAND RESPONSE

7.1 Introduction

TXU Energy launched the Brighten® iThermostat and Brighten® Conservation Program in June of 2008. The program included a Zigbee-enabled programmable thermostat that communicates to a Zigbee enabled gateway. The gateway was connected to the Internet via a customer-provided router to give customers a full range of remote thermostat controls and reporting, which could be accessed from most computers and smart phones. This program had two main purposes: 1) provide an incentive for recruiting and retaining customers in the highly competitive retail market, and 2) serve as a DR program that could respond to market needs and reduce load during critical periods.

The initial 2008 pilot started as a customer self-install model with the option for professional installation. Results of the self-install model quickly revealed that less than 35% of the devices came “online” primarily because customers didn’t bother to install the thermostat and/or the gateway. The program was then converted to a 100% professional installation model in 2009 where trained technicians installed the thermostat and gateway, then provisioned the equipment to ensure remote connectivity. By switching to a professional installation model TXU Energy was able to ensure initial connectivity; however, the percentage of online devices available for DR still remained well below anticipated levels. While some of the offline devices could be attributed to hardware issues, vendor quality assurance, and software issues, many were assumed to be the result of other customer-driven issues (disconnecting the gateway, Internet issues, etc.). As the customer base grew, so did the number of offline devices that were not able to receive a DR command.

While much of the DR market has focused on adjusting thermostat set points for a number of consecutive hours, TXU Energy implemented a DR program based on 15-minute cycling of air conditioning units. Based on a historical online percentage, for every 10,000 thermostats participating in the Brighten® Conservation Program, TXU Energy was only able to cycle approximately 5,500 air conditioners. Assuming each air conditioning system used an average of 1.5 kW, then the remaining 4,500 offline devices equated to approximately 6,750 kW of load that could not be reduced during a DR event. Since Texas has a Smart Meter Texas (SMT) portal that is capable of delivering DR signals to customer thermostats via the transmission and distribution utility (TDU) AMI network, the purpose of this project was to leverage that capability and provide two communication paths to the thermostat, one via the Internet and one via the AMI network. By utilizing both paths, TXU Energy expected to increase the percentage of controllable devices to over 95%, providing a better return on their investment in customer equipment.

The planned devices for this effort included:

- Gateway. The fielded Internet-capable configuration employed a gateway that transferred signals to and from the Internet to the thermostat. The concept was to pair (bind) that gateway to the residential smart meter so that it could receive signals over both communication paths.
- Programmable control thermostat (PCT). The PCT controls a home's heating and cooling system and it can be programmed to automatically change temperature at certain times of the day, generally to save energy. TXU Energy had deployed various models of PCTs in support of its iThermostat DR program.
- Smart meter. Texas has installed more than 6 million residential smart meters, and homes participating in this project are so equipped. For the initial demonstrations, this project planned to leverage the Landis & Gyr smart meters deployed in the Oncor Electric territory.
- Cable/DSL modem. These devices support the existing DR programs which use the Internet and provide the customers with the ability to control their PCTs and monitor various results.
- Smart Meter Texas portal (SMTP). This software system houses archived 15-minute increment data from each of the residential smart meters, and provides a means for customers to review their usage. It also allows residents to pair in-home devices to their smart meter. The system also supports the market with the capability to delivery DR signals to the smart meters, and subsequently any in-home devices paired to that individual meter.

The figure below illustrates the high-level proposed architecture for this project.

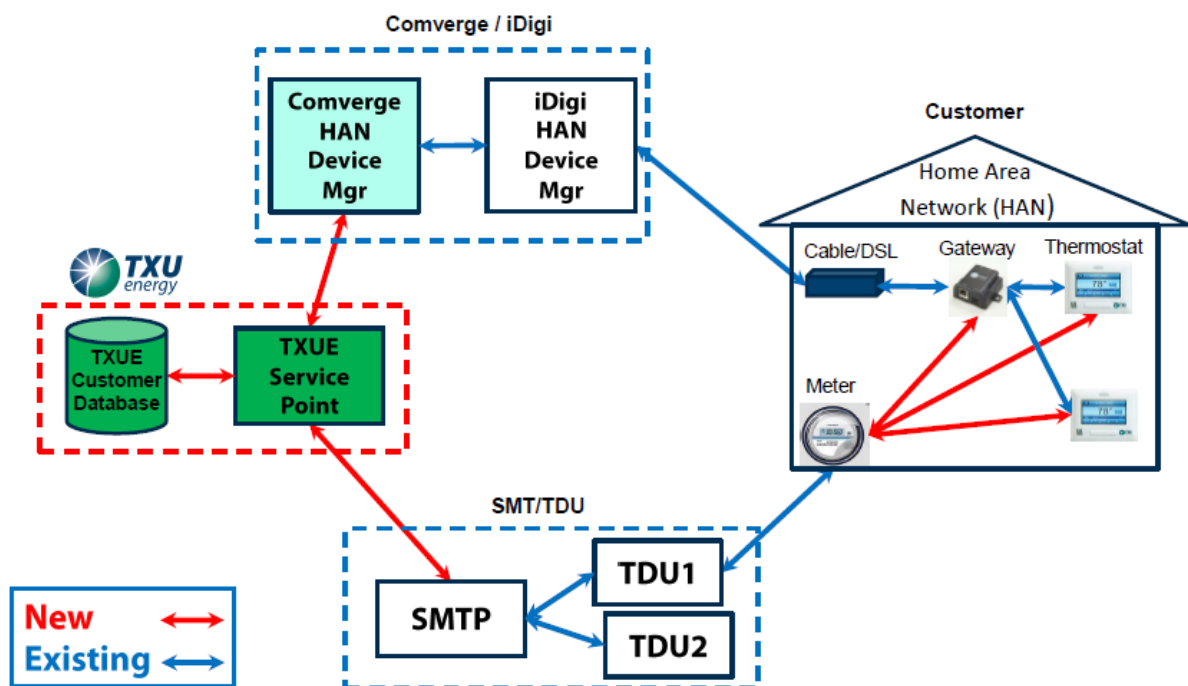


Figure 78. SMT Planning Project Architecture

7.2 Anticipated Benefits

Broadly, the expected benefits of the CCET demonstration project were to be derived from a more reliable electric grid that could facilitate effective management of and responses to increased wind resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage. Customers would be empowered to effectively and reliably manage their peak demand, therefore resulting in reduced customer electricity costs, reduced system-wide capacity needs, reduced electrical losses, and reduced environmental impacts. Specific smart grid benefits that were potentially supported by the residential DR component included:

- Improved DR results with more load shed during critical periods as a result of DR signals reaching more thermostats.
- A proven, viable means for delivering DR signals to more thermostats, especially those unreachable via the Internet or not equipped with a broadband connection, by utilizing dual communication paths.
- Greater return on investment for the SMT portal by providing REPs and other third parties with a viable means of reaching in-home devices via the utility AMI networks.

7.3 Interactions with Key Stakeholders

This project was sponsored by TXU Energy, a major REP in the Texas market, and the lessons learned were to be shared by all REPs. The device and dual communications path testing was supported by major TDUs and they were to be informed of the results. Improved DR results are of interest to ERCOT.

7.4 Technical Approach

This section provides a detailed description of the technical approach. The first part, Project Plan, includes illustrations of the system configurations, defines initial steps to achieve effective deployment, identifies the steps taken to demonstrate the technology (including test procedures), and outlines the approach to conduct analysis to assess performance. The second part, Data Collection and Benefits Analysis, defines the technical approach for conducting the benefits analysis, including the methodology and algorithms when appropriate.

7.4.1 Project Plan

This project was divided into two phases. The first phase focused on defining the appropriate architecture and the best equipment configuration to support the objectives, discovering any issues that may need to be resolved, and generally proving the concept. It involves experimentation and interoperability testing in a utility smart meter laboratory to determine what's feasible and how. The second phase was to focus on building and deploying the technologies to demonstrate the dual-path DR capability, including an application that leverages the SMT application programming interface (API) for

delivering DR signals via the AMI network. It would design and implement a solution. The objectives for these two phases are shown in the figure below, and the testing steps are defined after that.

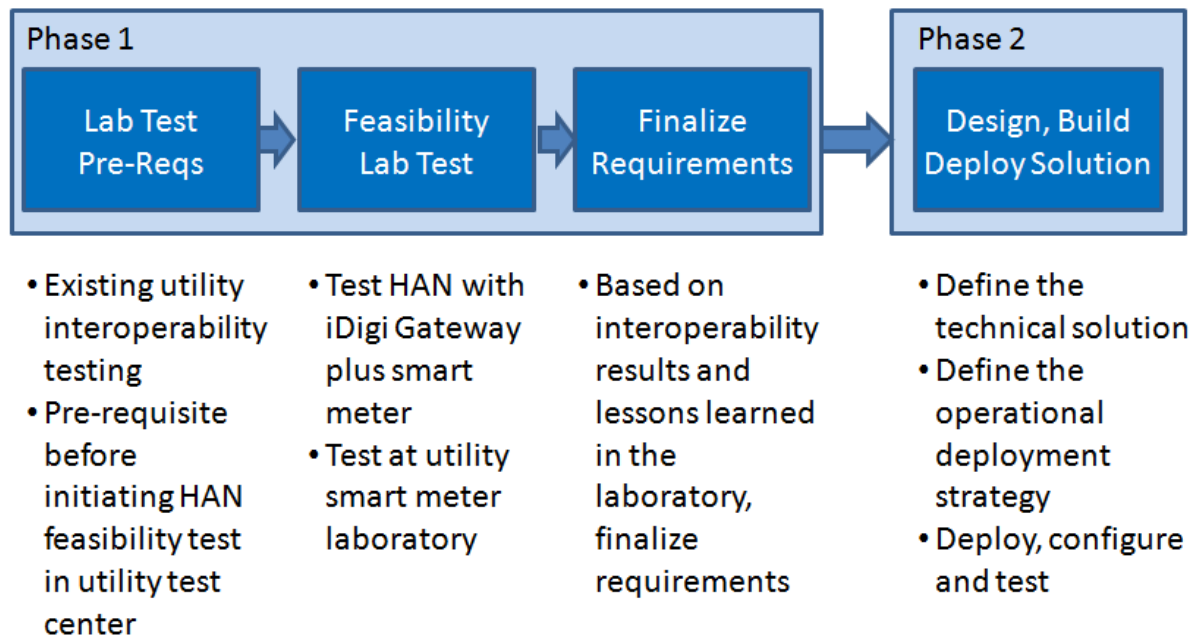


Figure 79. Phased Development of Dual Path Capability

Phase 1 – Laboratory Test at the HAN Test Laboratory

The objective of this test was to determine which configuration would support communication via both paths, the Internet and the SMT/AMI network. The scope of the testing involved the following:

- Initiated SMT commands from the Landis & Gyr test tool.
- Initiated vendor commands from their back-office system at the REP via Internet.
- Provisioned/De-provisioned the meter with the gateway (using SMT).
- Tested interoperability between the meter and the gateway.
- Tested interoperability from the meter to the PCT via the gateway.
- Verified ability to maintain connection with the vendor back-office system and the meter.
- Tested and discovered what happens at the PCT when it received the vendor proprietary DR command (via the Internet) as well as the SEP DR command (via the SMT).

Phase 1 – Feasibility Test at the Production Test Meter Laboratory

This second level of testing had three objectives:

- Determine if it was possible for a customer to switch out a gateway or would it require a truck roll with a trained installer.

- Become familiar (hands on) with the use of the SMT application programming interfaces (APIs).
- Verify work around for any identified vendor provisioning issue.

The scope of this testing included the following:

- Initiated SMT commands from SMT user interface SOAP tool into SMT test environment connected to the utility production test meter laboratory.
- Initiated vendor commands from back-office system via the Internet.
- Tested use cases of leveraging both communication paths to support the existing DR program. (Internet) and allow use of SMT communication path for DR commands.
- Tested work-arounds to vendor issue.
- Documented the combined installation/commissioning process and evaluated the potential for a customer self-install.

Phase 1 – Perform In-Home Test and Finalize Requirements

The objectives of the in-home test included:

- Determine if it is reasonable to expect a customer to self-install by switching out the gateway, provisioning the devices (if necessary), and contacting the vendor for verification.
- Work around any vendor DR issues.
- Work around any vendor global setting configurations from previous testing.

Based on the testing results, the Team determined the viability of proceeding to Phase 2. Although the approach was viable technically, it was determined that it would not be viable economically since the devices could not be updated remotely, and the customer self-install process proved to be problematic so it would require dispatching a team to install new equipment. Therefore, it was decided that the effort would not continue with Phase 2.

Phase 2 – Define, Deploy, Validate and Exercise the Dual Path DR Capability

Upon successful completion of Phase 1, and a GO decision, the Team would have conducted Phase 2. It would have:

- Defined the appropriate technical solution.
- Defined an operational deployment strategy.
- Developed an implementation schedule.
- Developed the software to interface with the SMT.
- Implemented the deployment strategy.
- Incrementally tested DR with both communications paths.
- Analyzed and documented DR results.

7.4.2 Data Collection and Benefits Analysis

TXU Energy already performs DR events to support grid stability. Once the team validated a workable solution for the dual path approach, the DR events would be called using both communication paths, and the results would be documented, compared and analyzed. In anticipation of Phase 2, and deployment/configuration of an appropriate dual-path solution, the team had defined its approach to reporting/validation as well as the methodology for data capture and analysis. Since Phase 2 did not occur, this portion of the reporting and analysis did not occur.

7.5 Performance Results

This section covers the operation of smart grid technologies and systems, including the demonstration results; impact metrics and benefits analyses; and any stakeholder feedback received to date.

7.5.1 Operation of Smart Grid Technologies and Systems

The current use of broadband networks to deliver DR signals to in-home devices has not achieved a high percentage of success, thus limiting the economic benefits of performing residential DR. The proposed solution was to also transmit the DR signals across the AMI networks, through the smart meter to the in-home thermostats. The Team anticipated technical hurdles that would need to be resolved so that the HAN devices could accept signals through two different communication networks. However, the Team anticipated that, once resolved, the improved success rate of these DR efforts could significantly contribute to reducing electric load across the grid when needed, and improve the economic viability of providing free thermostats to customers.

The Team began Phase 1 with the equipment configuration most widely deployed in the field, those being gateways configured as controllers paired to White Rodgers PCTs. Early testing and analysis revealed the following:

- The configuration of the White Rodgers PCTs supported a unique type of DR signal, but they were not smart energy profile (SEP) 1.x compliant so they were not configured to accept the DR signals that would be transmitted via the SMT.
- The existing White Rodgers PCTs were not configured to use production certificates which are required to pair with the smart meters.
- The gateways and PCTs would require software upgrades, and they were not remotely upgradable.
- The existing Digi x2 gateways were configured as ZigBee coordinators, and the PCTs could not pair with two coordinators, one being the gateway and one being the smart meter.
- The onboard memory for the PCTs was insufficient to add SEP text messaging.
- The cost to upgrade these devices would require a truck roll with a technician, and that cost was deemed infeasible. Estimates ranged from \$55 - \$180 per customer to roll a truck and

update/replace the gateways, depending on what was technically feasible. If new PCTs were required, this could cost an additional \$180 per customer.

The Team then conducted a workshop to evaluate alternative approaches to overcome the defined issues. They decided to repeat Phase 1 and continue with the dual path approach but switch configuration to a new product (Carrier PCT) that had been deployed by TXU Energy more recently. That approach is outlined below, and followed by results of the various testing.

HAN Test Laboratory

The objective of this test was to determine which configuration would support communication via both paths, the Internet and the SMT/AMI network. The scope of the testing involved the following:

- Initiated SMT commands from the Landis & Gyr test tool.
- Initiated vendor commands from their back-office system at the REP via Internet.
- Provisioned/De-provisioned the meter with the gateway (using SMT).
- Tested interoperability between the meter and the gateway.
- Tested interoperability from the meter to the PCT via the gateway.
- Verified ability to maintain connection with the vendor back-office system and the meter.
- Tested and discovered what happens at the PCT when it received the vendor proprietary DR command (via the Internet) as well as the SEP DR command (via the SMT).

The results of this testing revealed the following:

- Of eight possible combinations of gateway and HAN network configuration, one configuration accomplished a “partial pass”.
- One configuration supported the necessary communication via the new SMT/AMI path and provided partial communication via the existing vendor/Internet path.
- Defined three issues to review with the vendor to determine potential for full pass:
 - DR command via vendor system had an issue with the provisioning process—tested a work around
 - Telemetry data from the PCT to the vendor back-office system was not received. It is used for validation and also provides a customer-rich experience
 - Vendor framework global settings are based on gateway type. The router settings are different from the production controller environment settings
- Raised the question of whether or not the combined installation/commissioning process could be addressed by a customer performing a self-install to avoid the costly truck rolls.

Production Test Meter Laboratory

This second level of testing had three objectives:

- Determine if it is possible for a customer to swap out the gateway or if it will require a truck roll with a trained installer.
- Gain experience with the SMT HAN application programming interfaces.
- Verify work around for previously identified vendor provisioning issue.

The scope of this testing included the following:

- Sent various DR commands to the devices in the test laboratory using the SMT simple object access protocol (SOAP) tool interface.
- Initiated vendor commands from the back-office system via the Internet.
- Tested work-arounds to vendor issue.
- Documented the combined installation/commissioning process and evaluated the potential for a customer self install.

The findings of this testing were:

- Fully satisfied second objective. The team gained first-hand experience using SMT APIs.
- Partially satisfied first objective. The team evaluated approaches for a hybrid install process (call center supported installation process).
- Complete test could not be finalized. The vendor back-office test system experienced an outage.
- Team felt confident that all issues were resolved or could be worked around. They decided to repeat the test at the home of a “friendly” (TXU Energy employee), as outlined below.

In-Home Test

The objectives of the in-home test included:

- Determine if it is reasonable to expect a customer to self-install by swapping out the gateway, provisioning the devices (if necessary), and contacting the vendor for validation.
- Verify work around for vendor DR issue from previous testing.
- Verify work around for vendor global setting constraints from previous testing.

The findings of the testing included the following:

- The vendor back-office test system continued to experience challenges with full communications to the PCT. The vendor indicated that the optimal solution, and one that would be needed for actual operational implementation, would be to establish a second set of back-office systems focused on this new approach that uses gateways configured as routers. The estimated cost of that set-up was about \$400,000.
- The team was successfully able to bond the PCT to the smart meter using the SMT Portal.
- A DR signal as well as general message was sent to the PCT through the smart meter via the SMT Portal.

- Documented steps required to reconfigure the HAN in order to support the dual-path communication approach.

7.5.2 Metrics and Benefits Analysis

As part of this project, there were two key items that the Team focused on for analysis. The first was the general economic viability of residential DR in Texas. The second was the feasibility of using the SMT to provide a dual path communications solution that could improve DR results.

7.5.2.1 Operational Benefits

One of the key operational benefits explored is the economic viability of DR in the ERCOT market. Since 2008, TXU Energy has initiated DR events across 13 days, including five pre-season tests. TXU Energy currently deploys a 15-minute, 100% cycling strategy, and the customer topology allows for TDU segmentation as well as sustained cycling within a region by rotating customer groups across 15-minute intervals. This approach allows TXU Energy to respond quickly to initial grid conditions with a predictable load reduction, and also allows for flexibility when extended cycling is required.

In order to estimate the average customer load reduction achieved during a DR event, 15-minute customer meter data was pulled for all days where DR was executed in 2011 and 2013. The data included results for both cycled (online) and non-cycled (offline) customers. The two groups were then analyzed to:

- 1) Measure usage reduction for cycled customers.
- 2) Measure usage for non-cycled customer to establish a baseline trend.
- 3) Apply baseline trend to cycled customers.
- 4) Calculate difference between actual DR reduction and baseline for cycled customers.

The figures below reflect a portion of the actual DR results based on a sample size of ~6,000 total customers. The first figure shows the effects of a single DR event triggered at 4:55 p.m. (not fully aligned with the 15-minute interval data). The subsequent set of figures shows the effects of one or more events triggered over several days in August 2011.

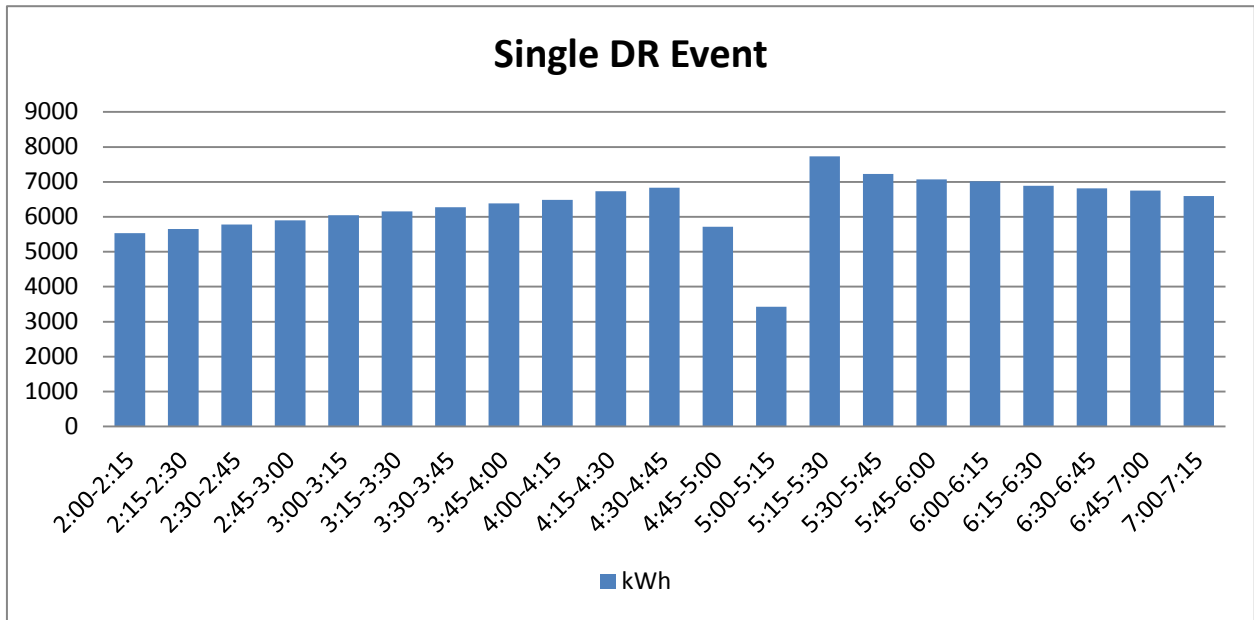


Figure 80. Single DR Event

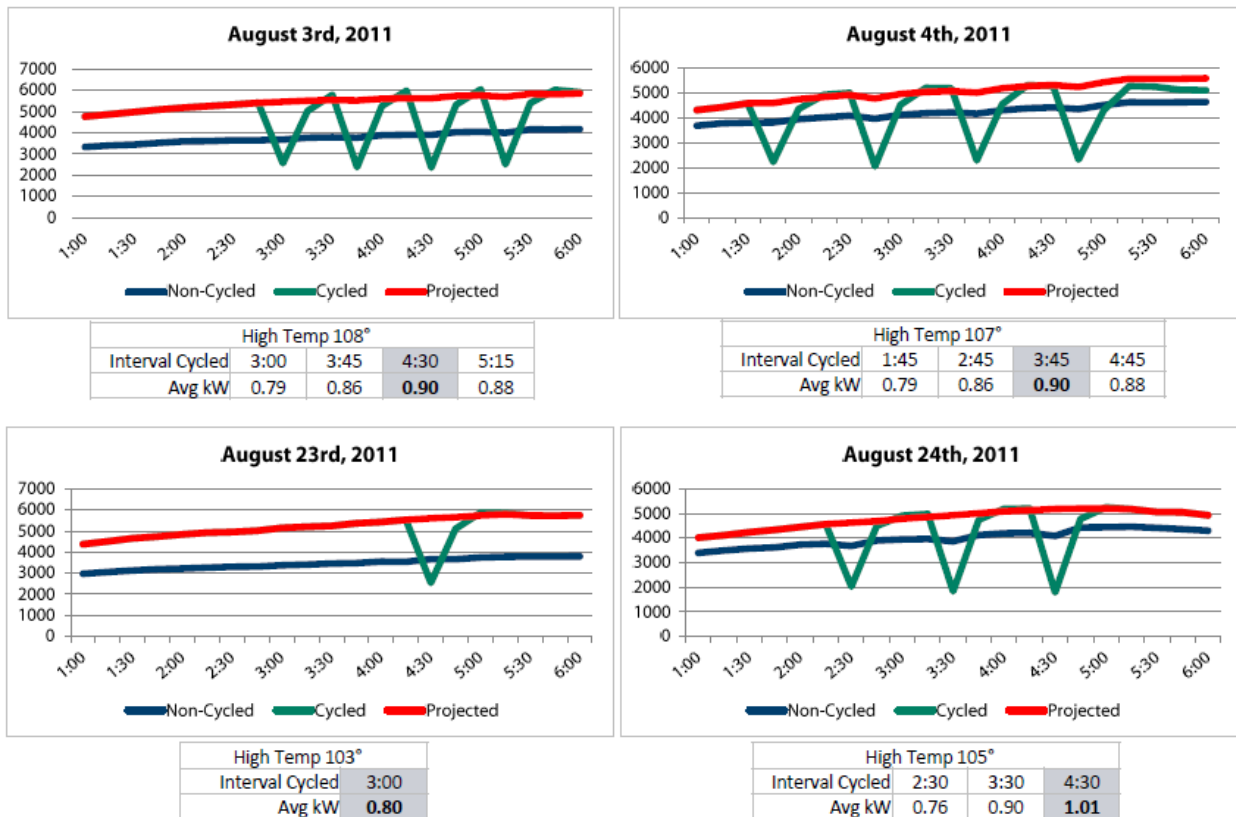


Figure 81. Multiple DR Events

Analysis of the 13 cycling days reveals the average reduction per customer (difference between cycled and smoothed) ranges from 0.50 kWh to 1.0 kWh. The variation can be directly correlated to outdoor temperatures. When cycling was performed on days where temperatures were in the upper 90s (not shown above), the average reductions ranged from 0.55 kWh to 0.75 kWh per customer. As temperatures climbed above 100 degrees the average reductions ranged from 0.75 kWh to 1.0 kWh per customer.

These averages can be used to determine the incremental value of controlling additional thermostats. From a pure cost perspective, the following chart represents the potential annual value of managing peak load at the maximum ERCOT bid cap levels. The table shows results for a suggested cycling of 1,000 homes for 15 minutes (maximum value per DR event) and then multiplies that by 10 DR events per year to derive the maximum annual value per 1,000 customers per year.

While these results can be extrapolated for a much larger DR pool of homes, the respective cost of equipment, truck rolls, and support must also be extrapolated. For example, in the table below, if a REP determined that they could outfit a customer with a gateway and PCT for \$300 (including equipment, truck roll cost, and ongoing support), it could take 6-10 years (depending on load shed amount) to recover that cost at the \$5,000 bid cap, and 4-6 years at the \$9,000 cap. Given that Texas is a competitive retail market, and customers can swap REPs at will, equipping them all with these devices with the primary intent of supporting DR would be a huge investment with equally great risk.

Table 29. Annual Value Proposition

Bid Cap (\$/MWh)	\$5,000	\$7,000	\$9,000	\$5,000	\$7,000	\$9,000
Less standard cost (\$/MWh)	\$90	\$90	\$90	\$90	\$90	\$90
Opportunity value (\$/MWh)	\$4,910	\$6,910	\$8,910	\$4,910	\$6,910	\$8,910
Average load shed per home for 15-minute event (kWh)	0.55	0.55	0.55	1.0	1.0	1.0
Customers cycled	1000	1000	1000	1000	1000	1000
Maximum value per DR event	\$2,701	\$3,801	\$4,901	\$4,910	\$6,910	\$8,910
# of DR events per year	10	10	10	10	10	10
Maximum annual value per 1,000 customers	\$27,005	\$38,005	\$49,005	\$49,100	\$69,100	\$89,100

A second key operational benefit explored is the feasibility of improving DR with SMT and dual-path communications. The testing demonstrated that it is technical feasible to perform DR using a dual communications path strategy. However, there are a number of technical issues that must be

addressed before it will be easily accomplished. Without operational trials, this project was unable to demonstrate the ability to improve DR results by sending the DR signals via two paths. One could estimate that the REP could potentially reach another 35% of their residents, meaning only 10% of the population was unreachable. However, as shown above in the economic analysis, the amount of time (in years) that it takes to recover the initial up-front investment makes it difficult to justify the costs simply to provide a DR capability. Certainly the dual path approach should be considered in defining a workable architecture that can support DR, and especially if the equipment investment is going to be made by the REP anyway. Such a solution could be an important contribution in future DR program approaches where the dual path capability adds only a small incremental cost to the set of items needed to equip and engage the customer.

7.5.2.2 Build Metrics and Benefits

Build metrics address equipment investments made for a project. For this DR effort, TXU Energy provided cost share in the amount of \$801,000 to cover the expenses related to installation and service, and some administrative support, for 1385 customers from 2011 to 2013.

7.5.2.3 Impact Metrics and Benefits

Impact metrics use reductions in energy use to compute reductions in respective environmental emissions. For the project, the calculation presumes a base of 6000 customers participating in 13 events with an average reduction in energy use of .75 kWh.

In computing pollutant emission reductions, the following formulas were used:

- CO₂: 0.00068956 tons/kWh
- SO_x: 0.00263084 kg/kWh
- NO_x: 0.00117934 kg/kWh
- PM_{2.5}: 0.001 kg/kWh

For the 13 project events, the resulting pollutant reductions were estimated to be:

- CO₂: 40.34 tons
- SO_x: 153.90 tons
- NO_x: 68.99 tons
- PM_{2.5}: 58.5 tons

If TXU were to continue and hold 10 DR events for a similar sized customer base and achieve similar energy use reductions, the annual pollutant emission reductions would be about:

- CO₂: 31.03 tons
- SO_x: 118.39 tons

- NOx: 53.07 tons
- PM2.5: 45 tons

7.5.3 Stakeholder Feedback

ERCOT and the TDUs are very supportive of DR efforts in Texas. TXU Energy has been able to support market demands and lower load when the grid was threatened with instability. It would be beneficial if all customers were enrolled in DR programs, but this would require a huge investment, and there is currently no way for REPs to easily and quickly recover the costs of such an investment.

Regarding the SMT, it is jointly funded by several of the TDUs and since it is capable of supporting DR, the TDUs would appreciate that service being fully exploited. Some REPs may explore this possibility, although it's more likely that third parties, such as home security companies, may look to leverage the ability to pair certain HAN devices to the residential smart meter and deliver DR or other services leveraging the capabilities within the SMT.

7.6 Conclusions

This section covers projections of demonstration and commercial scale system performance as well as lessons learned and best practices.

7.6.1 Projections of Demonstration and Commercial Scale System Performance

REPs like TXU Energy are already performing DR and equipping residential homes with devices that can support DR during critical times. With market incentives, these programs can be greatly expanded to not only broaden the number of homes with PCTs but also to extend the capabilities to other valuable in-home devices like water heaters, pool pumps, and electric vehicle charging stations.

For REPs with current DR programs leveraging PCTs or other in-home devices, they need to continue to explore alternatives to recover and maintain device connectivity. Multiple points of failure require a comprehensive strategy to identify and recover stranded devices:

- Proactive device monitoring and recovery to identify offline devices and implement remote recover processes.
- Proactive customer outreach with a comprehensive communication strategy (e-mail, outbound calling, direct mail, etc.) to trigger customer-initiated recovery.
- Customer surveys to identify customer characteristics for future targeting.
- Exploration of alternative hardware and network solutions.
- Analysis of the data and determination of optimal strategies for improved DR: logical groupings of homes that can sustain longer cycling or deeper setbacks, customers more willing to be inconvenienced when market conditions demand it, better ways to cycle or adjust thermostats without customer inconvenience that would support arbitrage, etc.

7.6.2 Lessons Learned and Best Practices

Although this project never achieved full operational capability, it did indicate a number of valuable lessons learned and best practices.

7.6.2.1 Lessons Learned

- There is a large number of early adopter PCTs in the field that are propriety in nature, and that are not SEP compliant. Even with newer PCTs that are SEP compliant, there still might be technical challenges in configuring them to support DR programs.
- Equipping a residential home with gateways, or even swapping out existing gateways, which is an item that typically plugs into an electric outlet, is a task that a homeowner can easily accomplish. With just a bit of instruction, they can easily walk through the process of pairing the gateways with other HAN devices, or even pairing them to the smart meter.
- Developing a dual path network will require a gateway configured as a router, not a controller, and the difference is not easily understood by the routine consumer.
- The differences in PCT configurations, and the fact that they are wall-mounted and connected to wires in the home, makes swapping out these devices much more of a challenge for homeowners.
- Leveraging the SMT to bond PCTs with smart meters and support DR programs will work for customers who don't have broadband service.
- Newly introduced PCTs are now SEP 1.x compliant, and some are even SEP 2.x compliant. Additionally, they are equipped with slots for two communications cards, so they could easily be configured to support both Wi-Fi and ZigBee, allowing for dual path communications and control.
- DR programs require SEP-capable in-home devices. Even the large box retailers are now selling PCTs that can support DR. Within Texas, the SMT provides a means for provisioning/bonding those devices to the smart meter so they can participate in DR programs. However, this will require that the REP educate the consumer, and that someone incent them to participate.
- Even at market caps of \$9,000, it will be very difficult for REPs in Texas to justify the expense of equipping a home with DR-enabled devices simply to perform DR. For greater market participation and penetration, new incentive programs will need to be established at the Federal or State level to incentivize introduction of these devices into more homes, and to make them available at critical times to support DR.

7.6.2.2 Best Practices

The resource investment in this project has resulted in key findings for TXU Energy that should be considered by stakeholders addressing the need for DR to contribute to ERCOT grid relief and in using SMT technology to do so. Those key findings include:

Customer Experience and Engagement:

- 1) The Texas electricity customer has an option of which provider they want to use to provide electricity to their home so the load reduction experience must be agreed to and present an opportunity for the customer.
- 2) A load reduction program should provide customers with options including:
 - a. Whether or not they choose to participate in any given event
 - b. Which thermostat they want to utilize (regardless of their REP)
 - c. Incentives to either reduce on their own or participate in a REP DR program

Cost Effectiveness:

- 1) Without TDU investment, REPs will not be able to effectively deliver a DR program as defined here and bear the full burden of the costs.
- 2) As shown in the previous Table, Annual Value Proposition, DR value does not cover the cost of a thermostat program as defined here either.
- 3) The market is evolving such that customers are willing to pay for a thermostat and installation (or self-install as that option becomes easier) thus sharing the cost burden.

Flexible Technology:

- 1) REPs, TDUs and hardware and software developers must be willing to develop technology that is flexible enough to interact with each other in order to make the most of their individual capabilities.
- 2) As Texas remains a competitive market, REPs will be encouraged to develop unique offerings with unique partners to maintain their competitive advantage which should further benefit the ERCOT market.

As the market evolves and new technologies are introduced to Texas consumers, REPS and TDUs will need to be prepared to respond in ways that address the above key findings to continue to be relevant players in this ecosystem. Energy efficiency is now a familiar term to Texas consumers, and they are beginning to understand the part that they play in helping the state. REPs and TDUs must provide the mechanisms through which they can participate and ultimately benefit the grid.

8. DISTRIBUTION-LEVEL BATTERY ENERGY STORAGE SYSTEM

8.1 Introduction

As part of the future community component, CCET was interested in investigating the benefits of deploying a battery energy storage system (BESS) in conjunction with wind turbines on a distribution-level grid. The overall objective was to test the ability of the BESS to perform various needed grid functions, such as peak shaving, load leveling, and frequency support. A secondary objective was to determine the economic viability of deploying these systems to support grid operations. The Reese Technology Center near Lubbock, Texas, afforded an optimal location for satisfying the objectives since it was already established and outfitted as a wind test facility, it provided access to a distribution grid operated by a local cooperative that was very supportive of testing, and was close to Texas Tech University which had a staff capable of performing the BESS evaluations. The Team defined a detailed functional specification for a BESS, and then issued a competitive request for proposal. After negotiations, CCET awarded a contract to Xtreme Power (now Younicos) for a 1MW Lithium Manganese Oxide (LMO) BESS. This system, pictured below, was built, tested, fielded, and connected to the grid in eight months, becoming fully operational in August 2013.



Figure 82. BESS Installed at the Reese Technology Center

The BESS houses 18 racks with 256 batteries totaling 4,608 battery cells with a combined energy storage capacity of 1 MWh.



Figure 83. Racks of Batteries within the BESS Container

The BESS operates in conjunction with about 4.6 MW of wind generation at the site. This includes an Alstom ECO86 1.67 MW wind turbine, a Gamesa 2 MW wind turbine, and three V27 300kW Vestas turbines that were deployed by Sandia National Laboratories as part of its new Scaled Wind Farm Technology (SWiFT) facility which currently produces 900 kW, and will eventually be expanded to produce 3.6 MW. The site also includes a 200-meter meteorological tower to provide atmospheric measurements for wind analysis. Two images of the SWiFT facility are shown below. The first figure below shows a view of the three SWiFT turbines and the meteorological tower. The second figure shows an aerial view of the SWiFT turbines and the close proximity of the BESS.



Figure 84. SWiFT Facility at Reese Technology Center



Figure 85. Aerial View of the BESS and the SWiFT Facility Turbines

The BESS is operationally integrated into the grid of the South Plains Electrical Cooperative (SPEC) and is controlled by SPEC with assistance from Younicos. During the course of the project, the BESS was thoroughly tested to determine its ability to support a number of functions and hence increase its value. Most importantly, it is being utilized to supply some of the energy necessary to meet the demands of a local substation electric load, especially during peak periods. For that purpose, it is typically charged during periods of low demand, which often coincide with high wind generation. However, since the battery system is connected to the utility grid, it does not need dedicated wind power to be charged.

The battery and wind generation are connected such that they are only used for local capacity and do not feed back onto the transmission grid beyond the boundaries of SPEC. The electrical system layout is shown below.

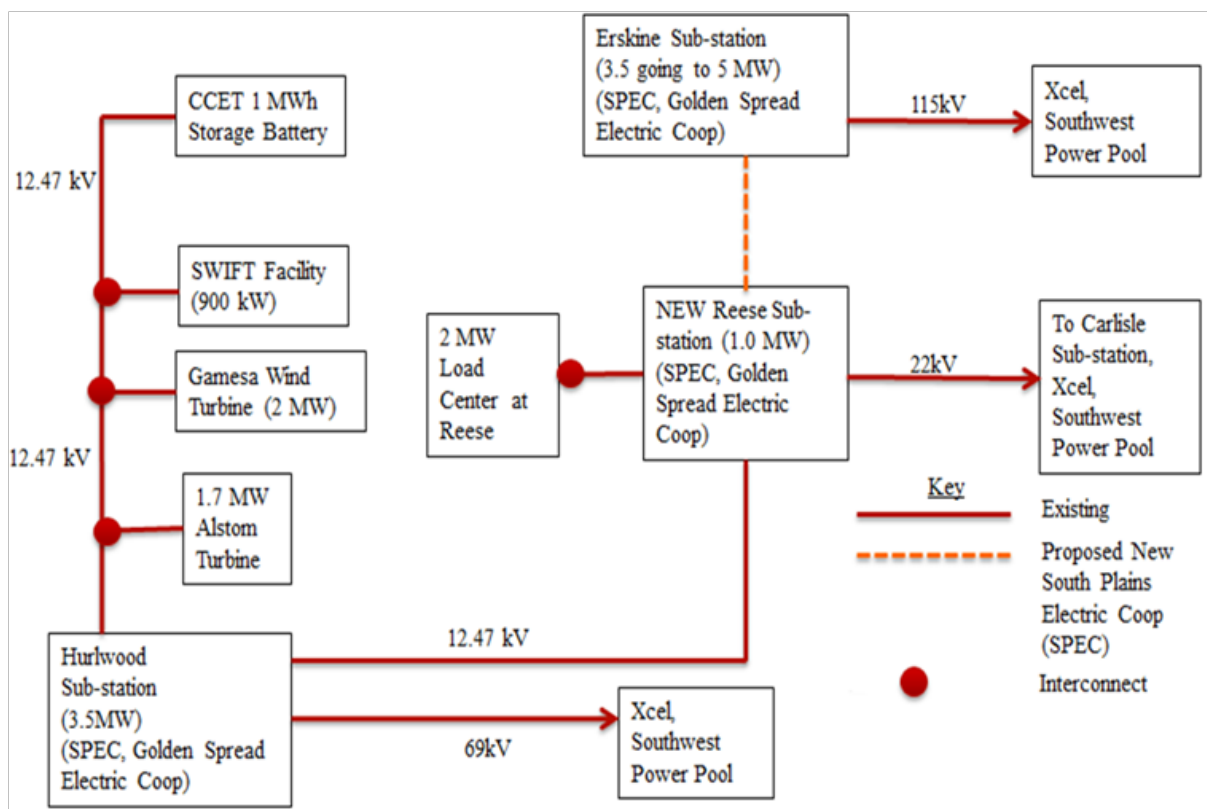


Figure 86. Reese Technology Center Electrical Layout

The BESS is connected via a bi-directional converter through a transformer to the main distribution grid. It is connected at the 12.47 kV medium voltage level, and can be charged or discharged by the bi-directional converter as shown below to manage power flow into the grid.

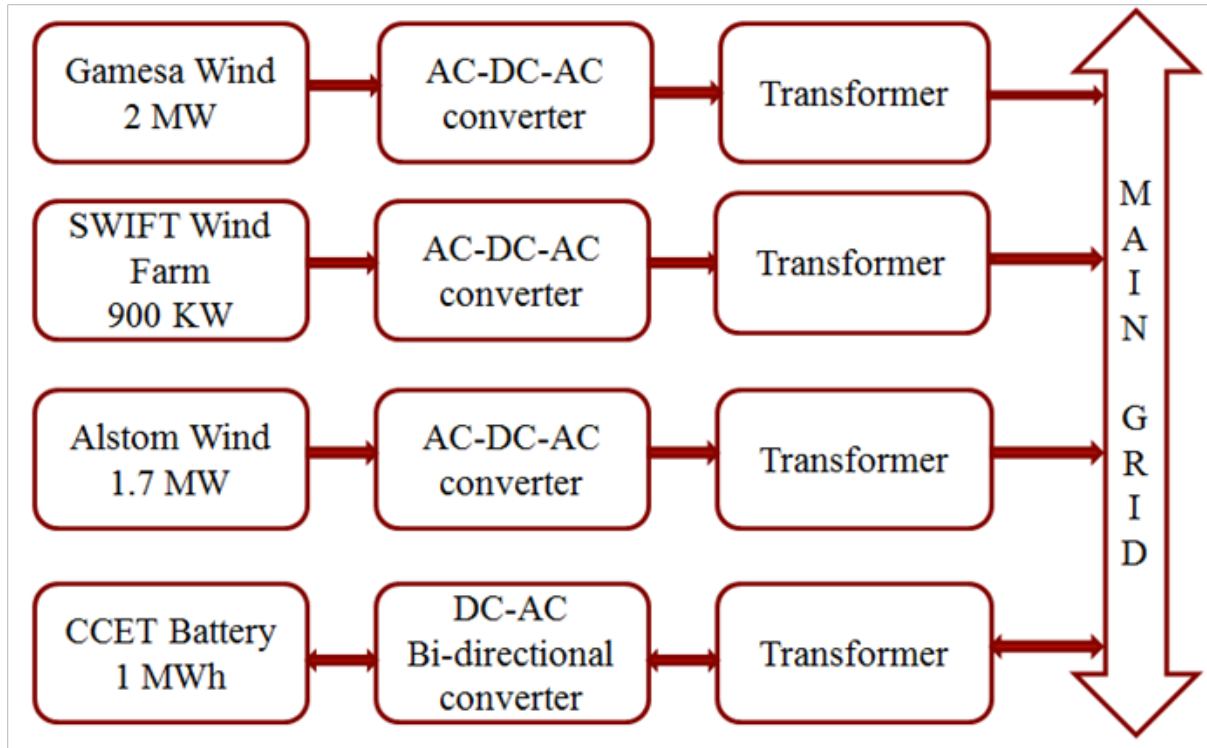


Figure 87. BESS Block Diagram

The BESS utilizes a series of deep-cycle LMO batteries developed by Samsung and designed for use in renewable energy applications. In addition to peak shaving for the local cooperative, SPEC, the BESS is also designed to perform other operations such as mitigating intermittent fluctuations of nearby wind turbines, regulating the distribution bus voltage through management of reactive power generation, serving as “spinning” reserve, and providing frequency support during the loss of generation.

8.2 Anticipated Benefits

Broadly, the expected benefits of the CCET demonstration project derive from a more reliable electric grid that can facilitate effective management of and responses to increased wind resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage. Customers are empowered to effectively and reliably manage their peak demand, therefore resulting in reduced customer electricity costs, reduced system-wide capacity needs, reduced electrical losses, and reduced environmental impacts. Specific smart grid benefits that were anticipated for the BESS included:

- Capture cheap wind power during periods of high renewable generation and low demand
- Feed real power into the SPEC distribution grid during peak demand periods
- Make renewable energy generation sources more dispatchable
- Make energy from renewable energy generation sources more valuable

- Serve as very fast frequency support to counteract loss of generation effects
- Serve as ramp support to offset fast decline of wind

8.3 Interactions with Key Stakeholders

The major interaction with stakeholders during the project was through the local utility, SPEC. TTU worked closely with SPEC to determine the potential roles of the BESS and how its performance could be maximized to help its customers. GSEC and SPP indicated an interest in the benefits derived from this system, and ERCOT and the CCET member utilities were extremely interested in discovering the value proposition for battery systems of this scale.

8.4 Technical Approach

This section provides a detailed description of the technical approach, the questions to be addressed, and the methods used to answer them. The first part, Project Plan, identifies the steps taken to demonstrate the technology, and outlines the approach to conduct analysis to assess performance. The second part, Data Collection and Benefits Analysis, defines the technical approach for conducting the benefits analysis, including the methodology and algorithms when appropriate.

8.4.1 Project Plan

The 1 MW BESS was installed at the Reese Technology Center to work in connection with 4.6 MW of wind generation at the site. The battery was utilized in conjunction with the wind facility to potentially supply all of the energy necessary to meet the electric load demands of the Hurlwood substation. The plan was to install the BESS and then evaluate its ability to support a wide array of functions that would improve its overall value for the grid.

The first step was to create very detailed and specific models for the battery, the wind turbines, and the utility network from the RTC to the substation in PSCAD to simulate the performance of the battery under several operating conditions such as:

- Offset load management – Charge the battery during periods of high wind and discharge the battery during the period of high load (peak load shaving).
- Ramp operation – Use the battery to supply energy when the wind speed drops rapidly, thereby countering the effective loss of the wind generation, providing time for other generation to be brought online.
- Frequency management – Discharge the battery at maximum rate to support the grid during frequency dips.

Once the system was modeled in PSCAD, the second step was to exercise the battery and perform a wide array of functional testing to determine potential uses for a BESS of this scale. These results were

then analyzed and used to validate the modeling results. Additionally, a third step was to perform an analysis to determine the economic viability for such a system.

8.4.2 Data Collection and Benefits Analysis

The technical approach included collection of a wide array of data including daily power production, wind speed and price data for the wind turbines, as well as the parametric data for the BESS. The latter included:

- Ramp rate and frequency for grid stabilization.
- Performance of the battery as a function of the SOC, temperature and the number of charge and discharge cycles. This also includes how well the energy flow is managed, in addition to how the transients and faults affect the battery.
- Response of the battery to steady state and transient requirements.
- Reliability of the battery as a function of performance.
- Parameters that affect the reliability of the battery such as over-current, over-voltage, under-voltage, and temperature.
- Power quality of the energy system at the connection point to the grid.

A complete analysis of power quality was performed on the output voltage from the battery, and the inverter connecting the battery to the grid. Because the battery was connected to the utility grid, transient analysis such as three-phase shorts, single-phase shorts and open circuit conditions could not be performed on the system. The transient analysis was, therefore, conducted by simulating the system and injecting the faults to determine how the components performed. Load flow and steady stated analysis were conducted to determine the optimal flow of power in the system.

8.5 Performance Results

This section covers the three main thrust areas: modeling and simulation, analysis of wind data, battery testing, and an economic analysis.

8.5.1 BESS Modeling and Simulation Efforts

Various types of generic models are available for the purpose of battery modeling, and these were investigated by the Team. A literature search indicated that the electrochemical polarization in Li-ion batteries leads to an inaccurate simulation of charge and discharge transients. Taking this factor into consideration, the Dual Polarization (DP) model was chosen for modeling the battery. This model has refined the description of polarization characteristics and simulated the concentration polarization and the electrochemical polarization separately. The resulting basic model for a battery cell is shown below.

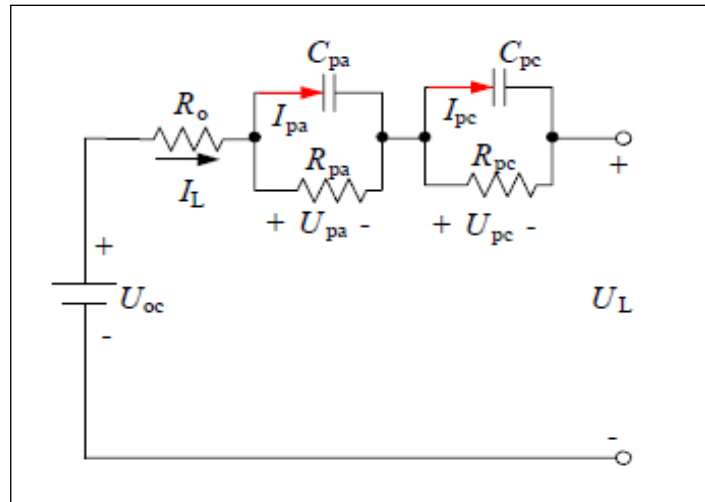


Figure 88. Basic Circuit of DP Model

The figure below shows the circuit with battery life-time and DP model ²⁶. This circuit has been modeled in PSCAD to obtain a 1 MW battery.

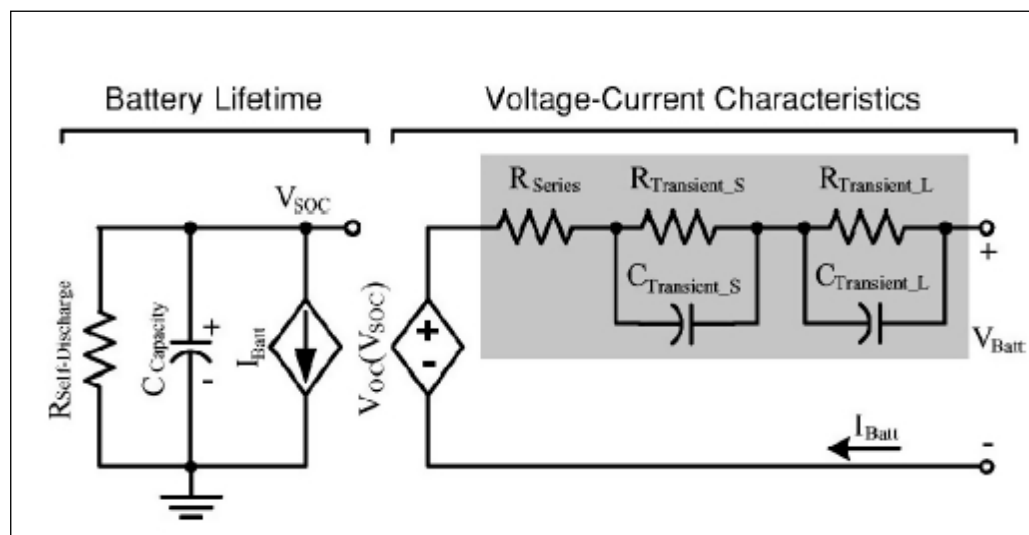


Figure 89. Accurate Battery Model

For the figure above, the characteristics are:

²⁶ Min Chen; Rincon-Mora, G.A, "Accurate electrical battery model capable of predicting runtime and I-V performance," Energy Conversion, IEEE Transactions on vol.21, no.2, pp.504,511, June 2006 doi: 10.1109/TEC.2006.874229

R_{series}	- Sudden voltage drop
$R_{transient_S}$ and $C_{transient_S}$	- First time constant
$R_{transient_L}$ and $C_{transient_L}$	- Second time constant
V_{batt}	- Battery terminal voltage
$C_{capacity}$	- Equivalent of energy stored in battery
V_{soc}	- Open-circuit voltage
$R_{self-discharge}$	- Ignored according to capacity retention

The discharge is controlled by current flowing out of the terminals of the battery. To discharge capacitor – I_{Batt} flows out through the current controlled current source. When the current flows out to load, it also discharges the capacitor, so the V_{soc} also starts to decrease since the V_{soc} controls the V_{oc} to control the terminal voltage V_{Batt} .

The charge current comes into the capacitor as reverse current flow. To charge the capacitor – I_{Batt} flows into the current controlled current source.

Based on the information above, the basic building block of a BESS, the modules, was formed by configuring the cells in series. Two system modules were then connected in series configuration and paired with tray-level battery management systems (BMSs) to form a rack-mountable system tray assembly.

8.5.1.1 Single Cell Battery Model

With reference to the above assembly and the calculations, initially a single cell of the battery system was modeled with the voltage at 4.12 V, a nominal voltage at 3 V and current at 60 A. The circuit is

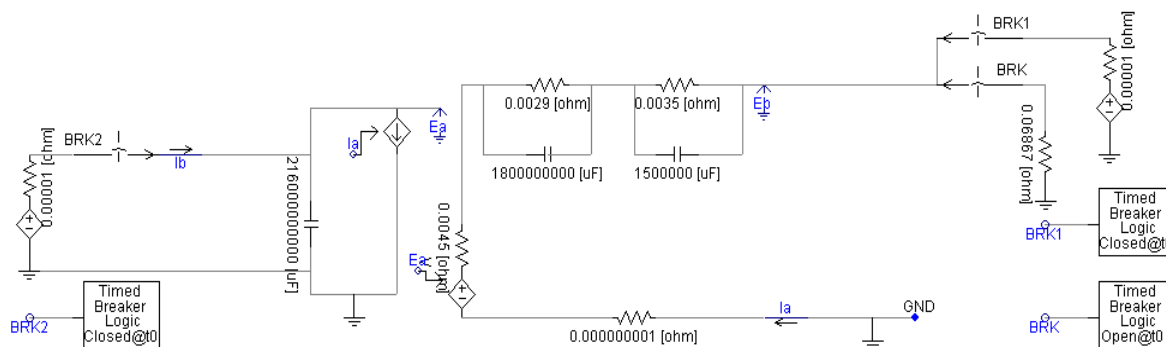


Figure 90. Single Cell Battery Model

shown below.

The figure below shows the controls to enable the battery roll-off before it drops further in reaching the lowest operating point.

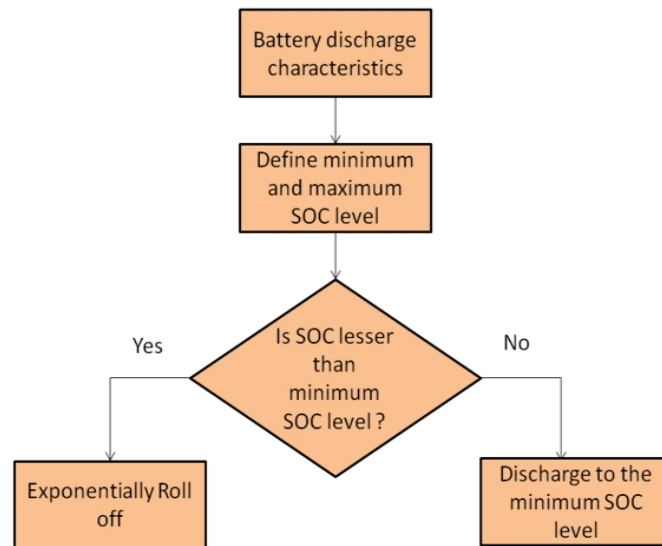


Figure 91. Control Algorithm of the Battery Cell

Based on the control algorithm above, the battery was charged to its full capacity of 4.12 V through the direct current (DC) voltage source. The battery was made to discharge at a 1 C rate (60 Ah) through a resistive load of 0.0689 ohm until it reached the minimum operating voltage of 3.0 V at which time the battery was made to exponentially roll off to zero (the line indicating the steady rate of increase or decrease levels off—basically curves away at an exponential rate). The battery voltage was observed as the battery discharged over time. Once the minimum operating voltage was reached, the comparator enabled the battery voltage ' E_b ' to roll off at 3600 seconds. The figure below shows the implementation of the controls as they were modeled in PSCAD. The voltage source at the right side of the model was used to charge the battery with 4.12 V during which the BRK1 breaker was closed and the BRK breaker was open. After the charging of battery, the BRK1 was opened and the BRK was closed so that the battery discharged through the load of 0.06867 ohm. The current ' I_a ' was used to charge or discharge the battery C_{capacity} dynamically. The output from the comparator was fed through a parallel connection of voltage source and a resistor of 0.6867 ohm to observe and plot the roll off at 3600 seconds.

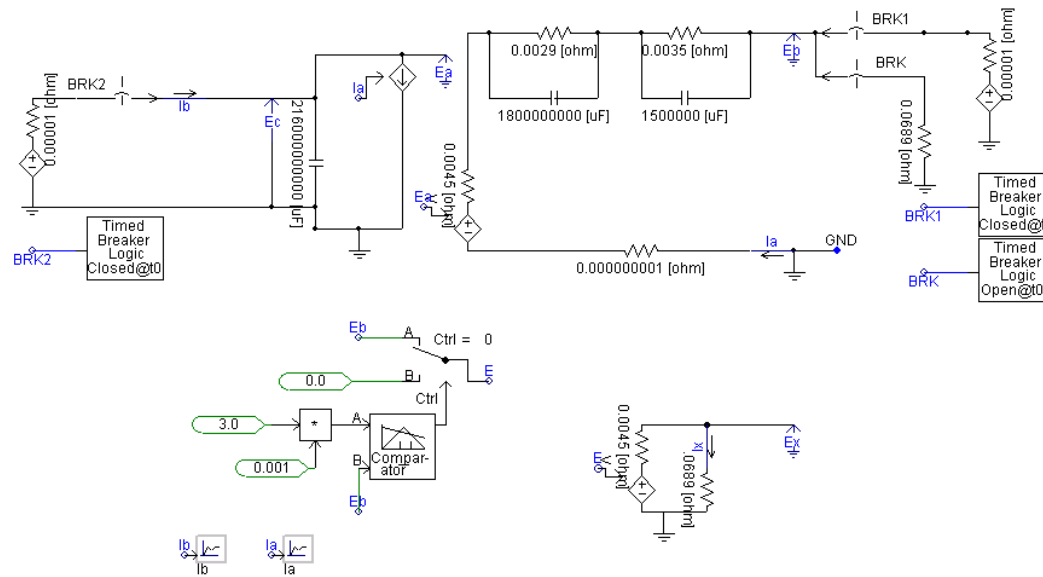


Figure 92. Model of the Single Cell Battery with Controls

8.5.1.2 Discharge Characteristics of Various C Rates

A C-rate is a measure of the rate at which a battery is discharged relative to its maximum capacity. The battery (1 MWh) operates until the minimum SOC level (760 V) at 1C which means the battery discharges at a constant current of 1200 A until it reaches minimum state of charge (SOC) of the battery in 1 hour. If the 1200 Ah battery is charged/discharged at a current of 600 A in 2 hours then it is called C/2 rate. If the 1200 Ah battery is charged/discharged at a current of 1200 A in 1 hour it is called 1C rate. If the 1200 Ah battery is charged/discharged at a current of 2400 A in 1/2 hour it is called 2C. The simulation results are shown below for various battery C rates given the 4.12 V, 60 Ah battery.

The battery discharge characteristics for 1C are shown below. The battery voltage was at 4.12 V initially and discharged up to the lowest operating voltage of 3.0 V for one hour (3600 seconds) at a constant rate of 60 A.

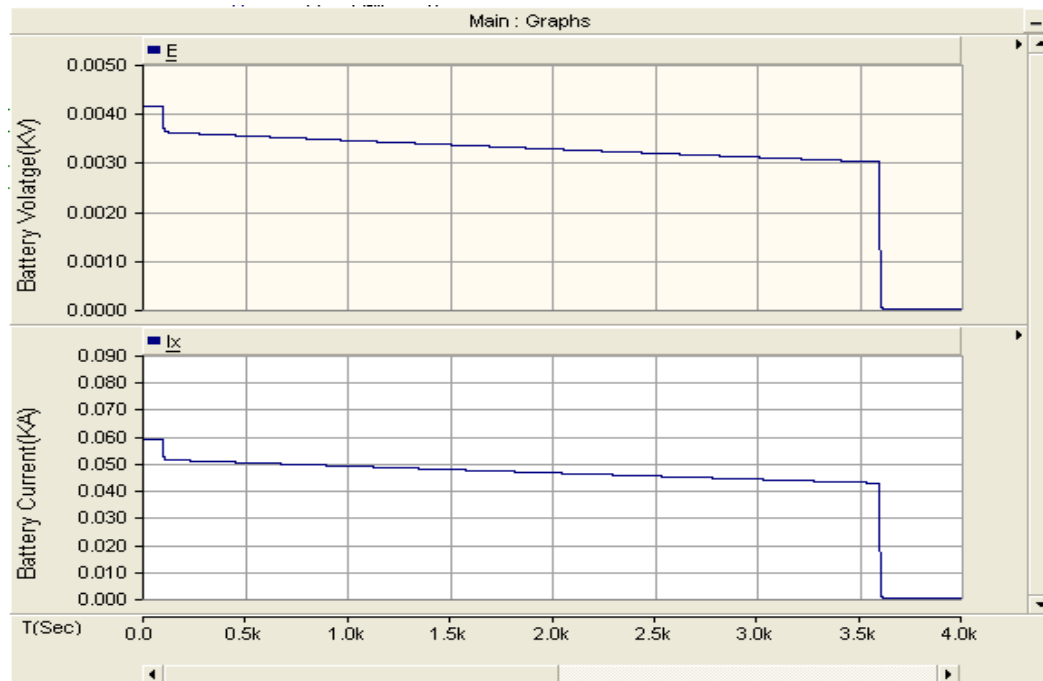


Figure 93. Battery Discharge at 1C rate

Similarly the battery discharge characteristics for C/2 battery are shown below. The battery voltage was at 4.12 V initially and discharged up to the lowest operating voltage of 3.0 V for two hours (7200 seconds) at a constant rate of approximately 30 A.

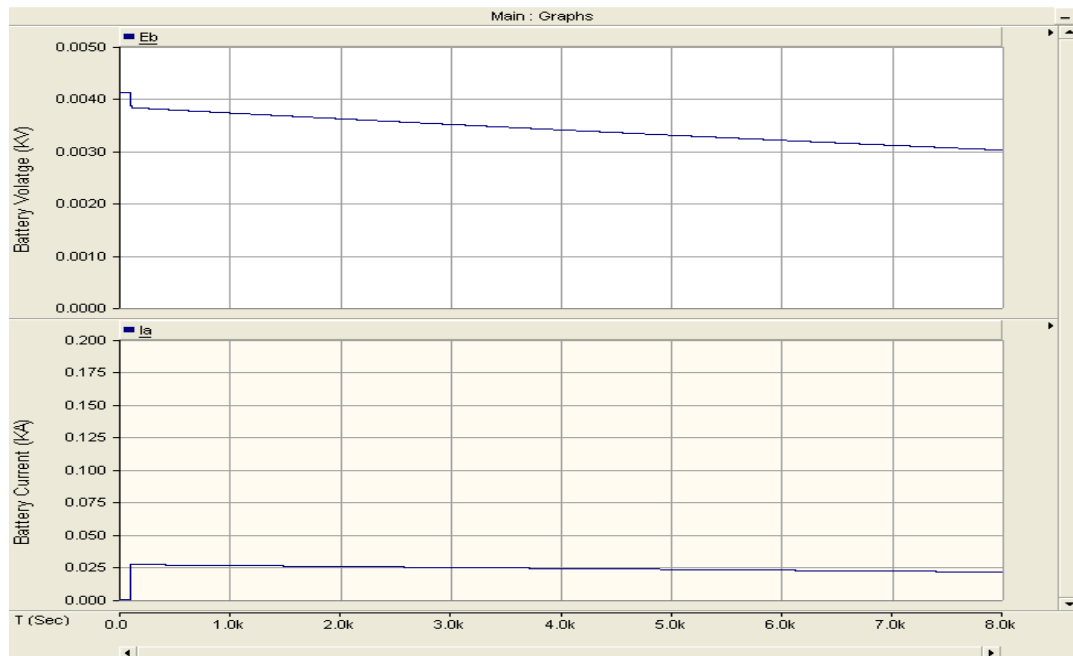


Figure 94. Battery Discharge at C/2 Rate

The battery discharge characteristics for 2C are shown below. The battery voltage was at 4.12 V initially and discharged up to the lowest operating voltage of 3.0 V for half an hour (1800 seconds) at a constant rate of approximately 90 A.

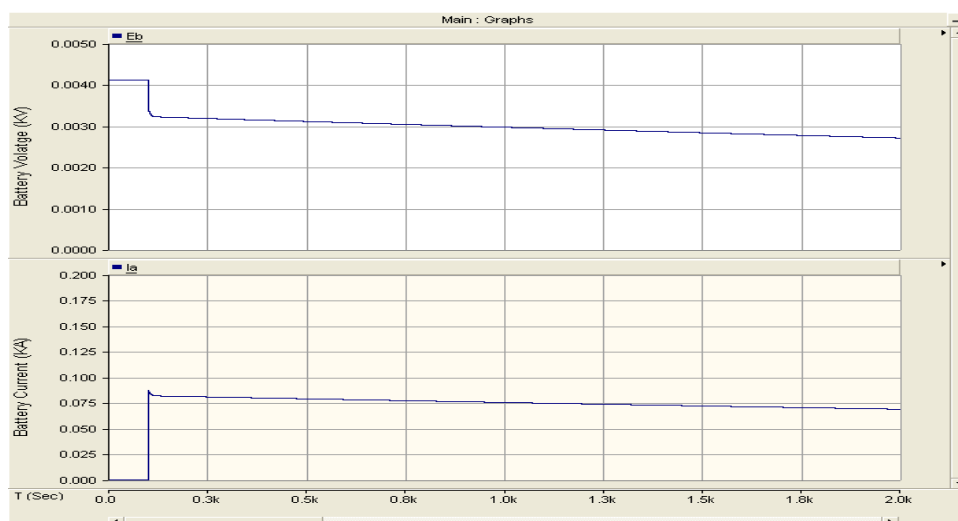


Figure 95. Battery Discharge at 2C Rate

Similarly the battery discharge characteristics for 3C are shown below. The battery voltage was at 4.12 V initially and discharged up to the lowest operating voltage of 3.0 V for 15 minutes (900 seconds) at a constant rate of approximately 110 A.

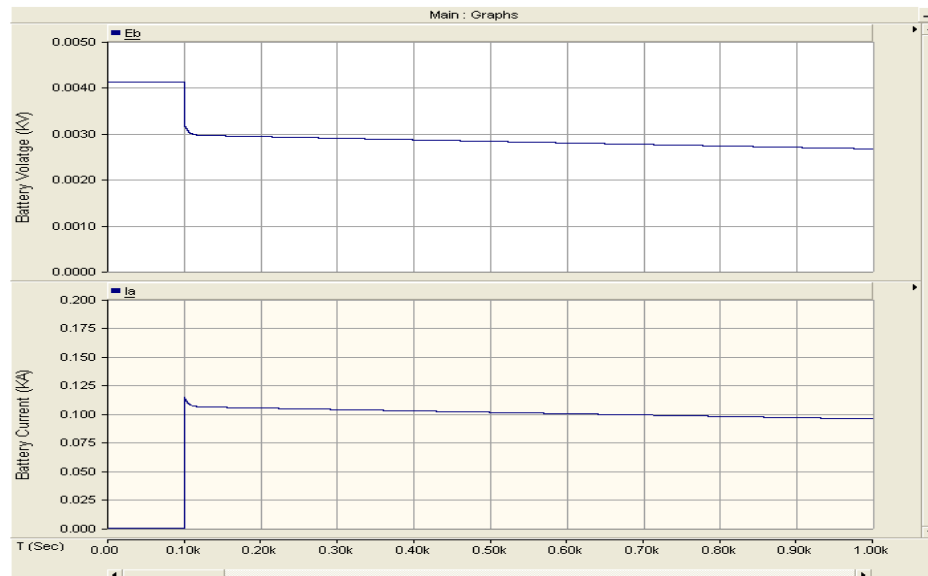


Figure 96. Battery Discharge at 3C Rate

The above simulation results are tabulated as shown below:

Table 30. Results for Different C Rates

Battery Cell Parameter	Cell	Simulation Results			
		C/2 (30 A)	1C (60 Ah)	2C (120 A)	3C (180Ah)
Nominal Voltage	3.7 V	3.7 V	3.7 V	3.6 V	3.6 V
Operating Voltage	3.0-4.12	3.0-4.12	2.9 – 4.12	2.8 – 4.12	2.8 – 4.12
Design Capacity	60 Ah	60 Ah	60 Ah	60 Ah	60 Ah
Battery Current	60 A	28 A	50 A	90 A	120 A

8.5.1.3 Validation of the Battery Modeling

The figure below shows the simulation data compared to the experimental cell data of a 4.12 V, 60 Ah LMO single cell. The simulation of the single cell battery model in PSCAD was saved to an output file. Using the MATLAB program, the simulation data of battery voltage was read from PSCAD and the experimental data was read from an Excel file. The graphs were plotted and the slopes were calculated

as 0.0002 for experimental and 0.000196 for simulated data respectively. The percentage of error in the model is found to be 0.2% and is calculated as follows²⁷:

$$\% \text{ of error} = [(\text{Simulated slope} - \text{experimental slope}) / \text{experimental slope}] * 100$$

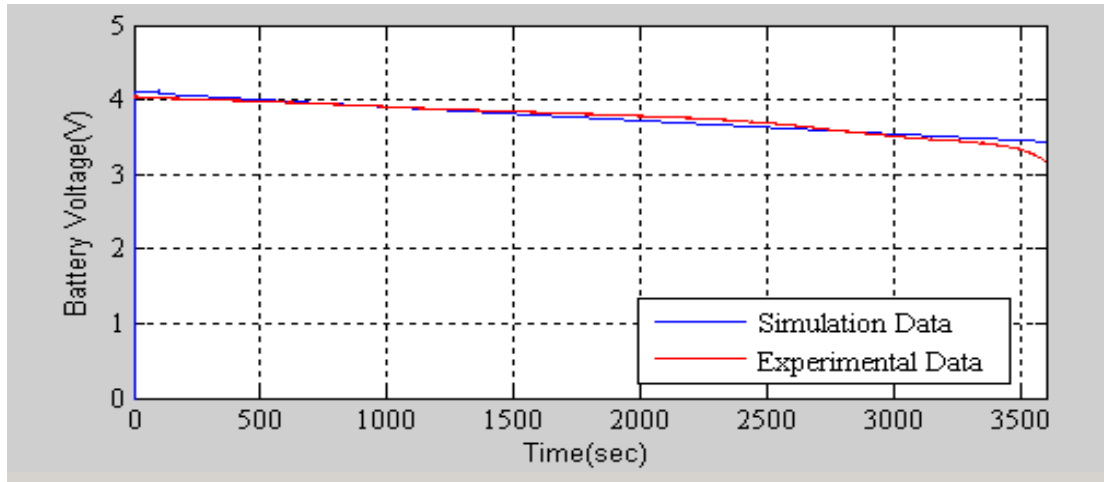


Figure 97. Comparison of Simulation and Experimental Results of 1C (60 A)

8.5.1.4 Stacked Cell Battery Model

The battery terminal voltage response includes voltage drop with a sudden and a slow response curve and hence the battery terminal voltage response is described in the short and long term operating conditions by the RC network. The basic building blocks of the BESS, modules, were formed by configuring the cells in series. To obtain a 1 MWh battery, an operating voltage of 760 V to 1053 V with a nominal operating voltage of 947 V and 1200 Ah was considered. The battery was charged to 4.12 V and discharged slowly to reach the nominal voltage of 3.7 V at 3600 seconds. The 16 single cells were stacked in series, and further stacked up to get 256 cells (256 cells * 4.12 V = 1055 V) with a voltage of 1055 V.

Once the single cell of 3.7 V was modeled, 256 such cells were stacked in series to obtain a voltage of 1055 V and current of 60 A. Modules were connected in parallel to increase the current to 1200 Ah with 1055 V and resistive load of 0.78916 ohm. The battery was charged to 1055 V and made to discharge slowly and made to exponentially roll off at 3600 seconds as shown below.

²⁷ Broughton, Kathleen. "A Tale of Two Processes: Simulation vs Experimentation." University of Illinois, Chicago, 2011. Web. 23 Sept. 2014. <http://www.slideshare.net/broughtonkm/simulation-vs-experimentation-a-tale-of-two-processes>

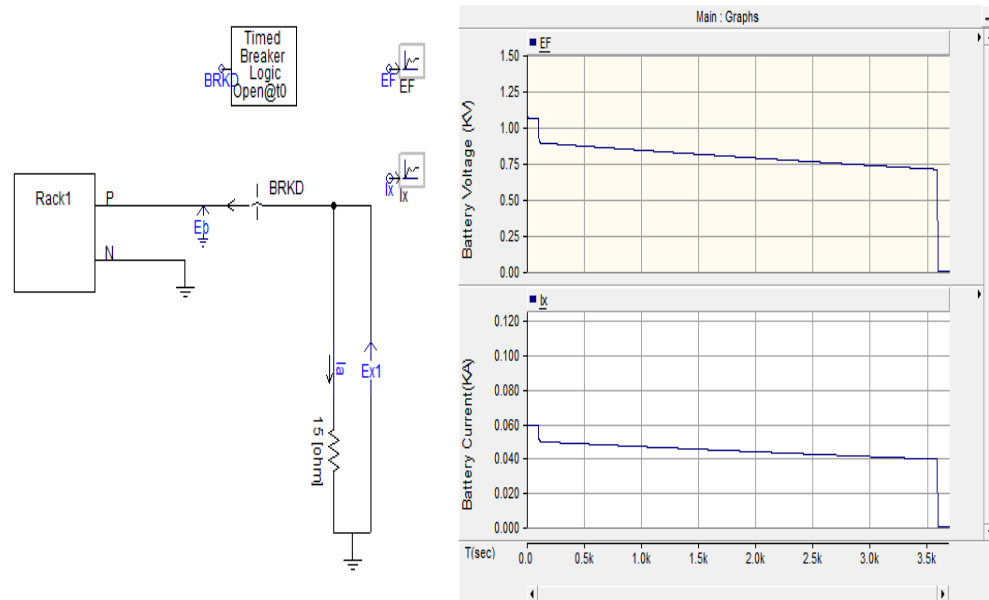


Figure 98. Voltage and Current (V & I) Waveforms of the 1055 V, 60 Ah Battery System

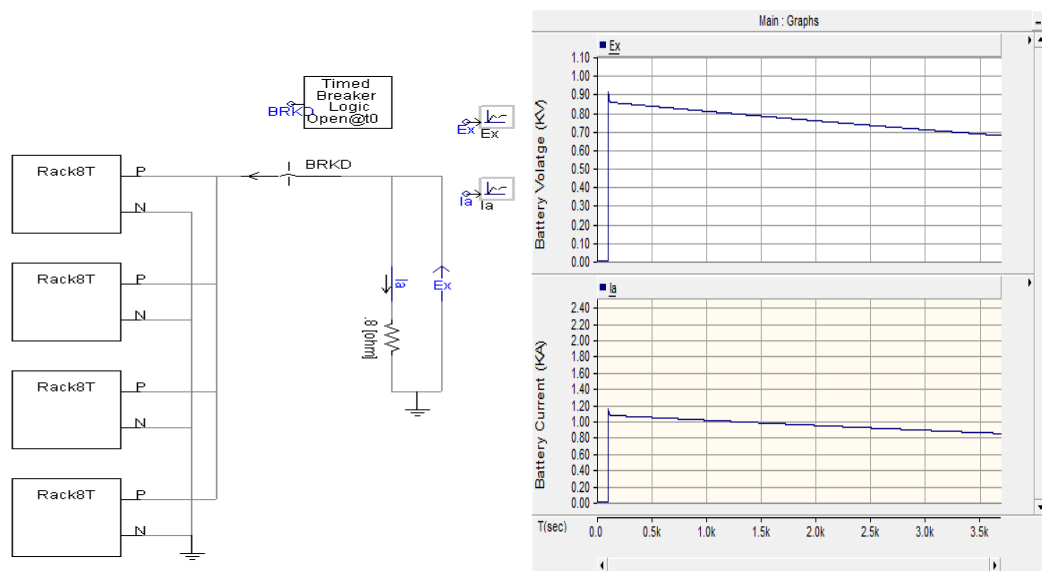


Figure 99. V & I Waveforms of the 1055 V, 1200 Ah Battery System

8.5.1.5 Electrical Equivalent Cell Model

Equivalent circuit models are often used to simplify the calculation whereby a complex circuit is represented in its simplest form to enable analysis of the circuit models. The equivalent circuits describe the electrical behavior of the model. Also issues with node compatibility, accuracy, memory space requirements and run time of a particular program in PSCAD are overcome by using the electrical circuit model.

The expression for the controlled voltage source is given by the following equation²⁸ with E_0 - battery constant voltage, E - no-load battery voltage, K - polarization voltage, Q - battery capacity (Ah), $\int i dt$ - actual battery charge, A - exponential zone amplitude, and B - exponential zone time inverse.

$$E = E_0 - K \frac{Q}{Q - \int i dt} + A \exp(-B \times \int i dt)$$

Instead of stacking the cells, the equivalent circuit for the 1 MWh battery was modeled and analyzed for the discharge characteristics of a resistive load. The calculation of parameters of the electrical equivalent cell was similar to the procedure of a single cell. Assuming battery was at charged state, the usable capacity can be modeled by a full capacity capacitor C_{capacity} . By setting the initial value to be 1200 V, the battery was initially fully charged (SOC is 100%). C capacity will not change with current variation. When the battery was charged or discharged, the controlled current course, I_{Batt} was used to charge or discharge C_{capacity} so that the SOC represented by Voltage SOC (VSOC) changed dynamically. Hence the battery run time or the discharge slope was obtained until the battery voltage reached the end of discharge or the minimum operating voltage. The battery was charged to 1053 V and made to discharge slowly and made to roll off at 3600 seconds at 1C rate as shown by battery voltage, E_b , in the figure below. The current across the resistive load was also found to be discharging slowly from 1200 Ah as

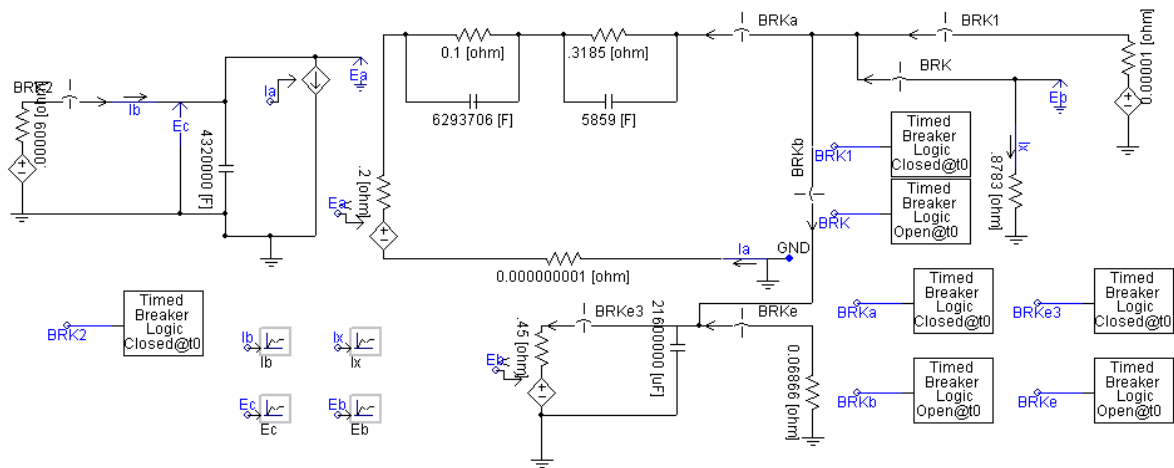


Figure 100. Equivalent Model of the 1MWH Battery

²⁸ Ali, S. M., Pattanayal, P., Sanki, P., & Sabat, R. R. (2012). Modeling and control of grid-connected hybrid photovoltaic distributed generation system. *International Journal of Engineering Sciences & Emerging Technologies*, 4(1), 125-132.

shown by battery current, I_x , in the second figure below.

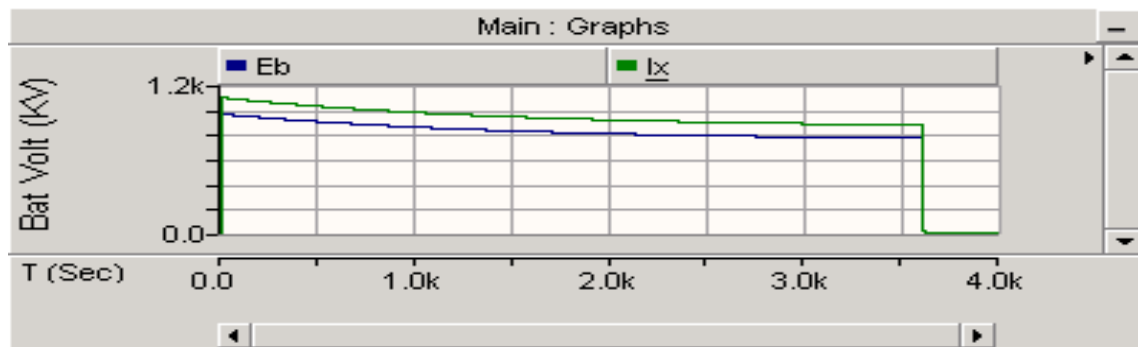


Figure 101. V & I Waveforms of the 1 MW Battery System

8.5.1.6 Single and Stacked Cell Battery System Comparison

The two types of battery models, such as “stacked cell model” and “single cell model”, of a 1 MWh LMO battery were taken into consideration to study the discharge characteristics. The approach was based on the use of the Dual Polarization Model (one of the Electrical Equivalent Circuit Models also called Two Time Constant Model). The purpose was to build and simulate the battery models in PSCAD. Also the battery models were compared and analyzed on their performance during faults when connected to the resistive load at 1C discharge rate.

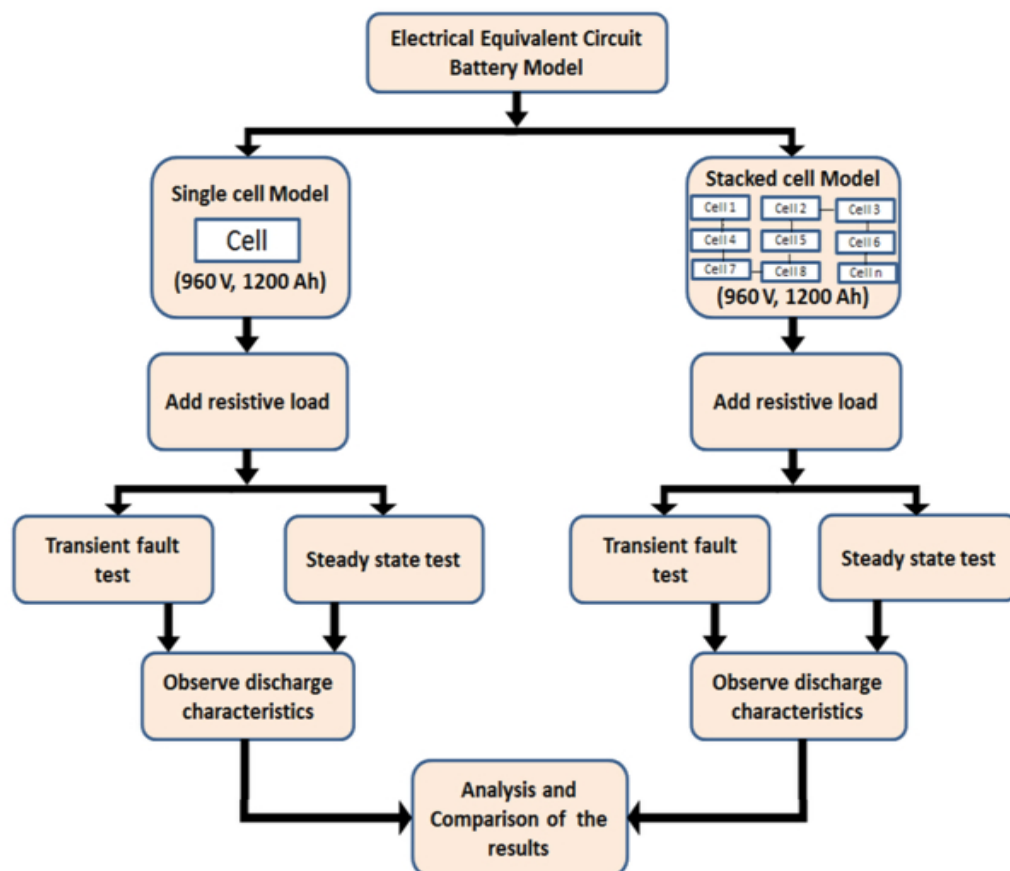


Figure 102. Process Flow for Analysis of Battery Modeling

The figure above describes the process flow analysis of the battery modeling to analyze the models in terms of discharge characteristics, speed, performance, stability and compatibility. The “single cell model” was constructed using the electrical equivalent circuit. In an attempt to simulate the “stacked cell model”, initially one cell of the battery rated at 4.12 V, 60 Ah was modeled in PSCAD and then the cells were connected in series and parallel to build the model of 1 MWh battery module. The key results include the modeling of single cell and stacked cell models of the 1 MWh battery. The discharge characteristics of the battery models rated at 960 V, 1200 Ah were analyzed for the faults applied near the resistive load. The models were tested for node compatibility, compilation issues and the response effectiveness when connected to the resistive load.

8.5.1.6.1 Single Cell Discharge Characteristics

In order to compare the performance of the stacked cell model with the single cell model, the stacked cell model was built in PSCAD. Once the single cell of 4.12 V, 60 Ah was modeled, the electrical equivalent single cell was modeled for 1055 V and 1200 Ah using the equations of dual polarization. The models were analyzed for the battery discharge characteristics for the resistive load as shown below. The voltage source at the right side of the model was used to charge the battery with 1100 V during which the BRK1 breaker was closed and the BRK breaker was open. After the charging of the battery, the BRK1 was opened and the BRK was closed so that the battery discharged through the load of 0.8775 ohm. The current “ I_a ” was used to charge or discharge the battery $C_{capacity}$ dynamically. The output from the comparator was fed through a parallel connection of voltage source and a resistor of 0.8775 ohm to observe and plot the roll off at 3600 seconds.

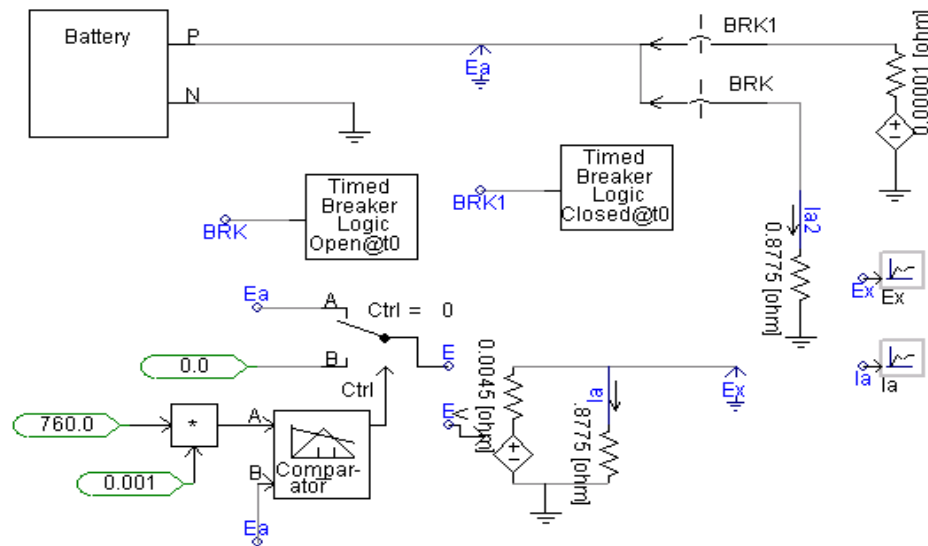


Figure 103. 1 MWh Battery Equivalent Single Cell Model in PSCAD with Controls

The open circuit voltage was set to the maximum voltage as 1053 V. The battery was charged to 1053 V and made to discharge slowly and made to roll off at 3600 seconds at 1C rate as shown by battery

voltage, “Ex”, as shown in the figure below. The resistive load was calculated to be 0.8775 ohm for the current to discharge at a constant rate of 1200 Ah as shown by battery current, “I_a”.

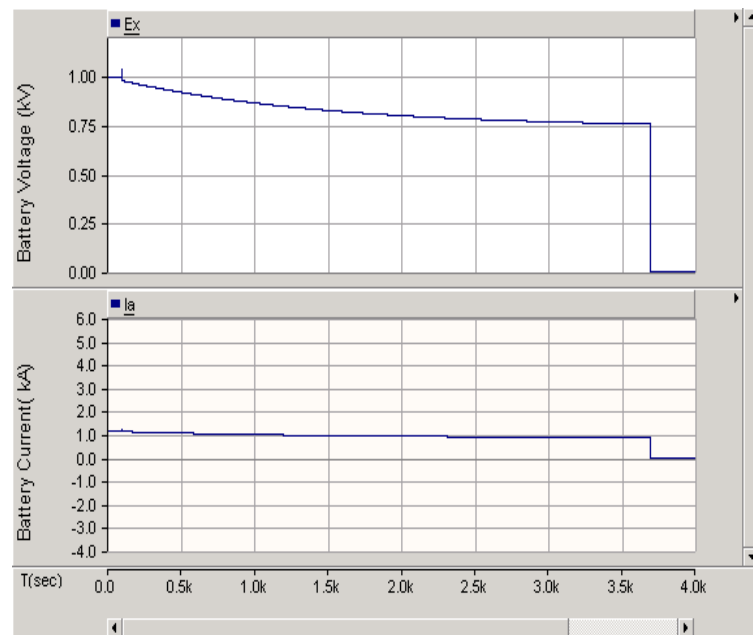


Figure 104. Voltage & Current Waveforms of the 1 MWh Battery

8.5.1.6.2 Stacked Cell Discharge Characteristics

The stacked cell model system with controls and the obtained discharge curve is shown below.

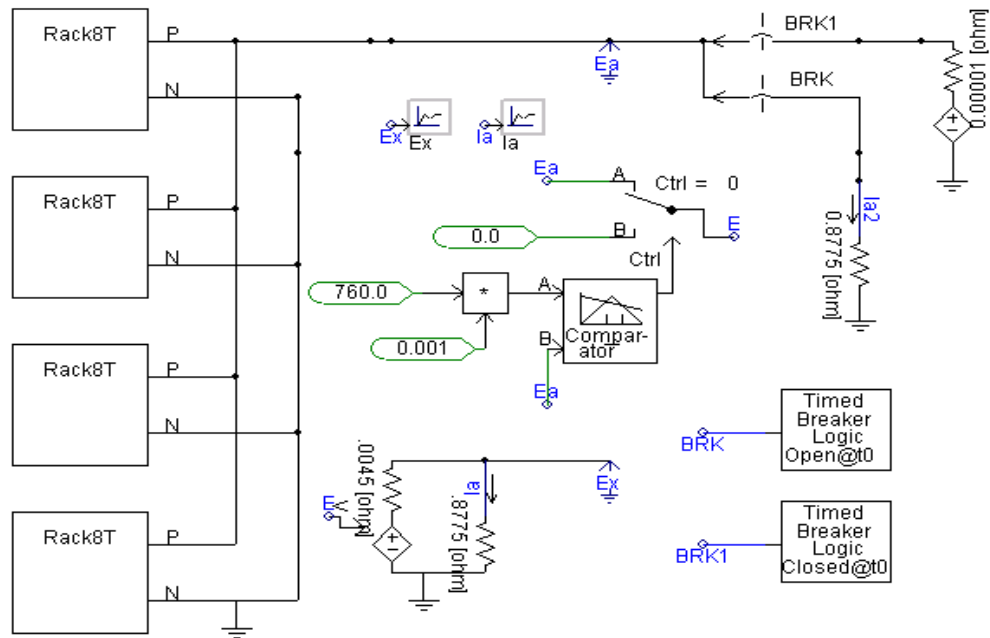


Figure 105. 1 MWh Battery Stacked Cell Model in PSCAD with Controls

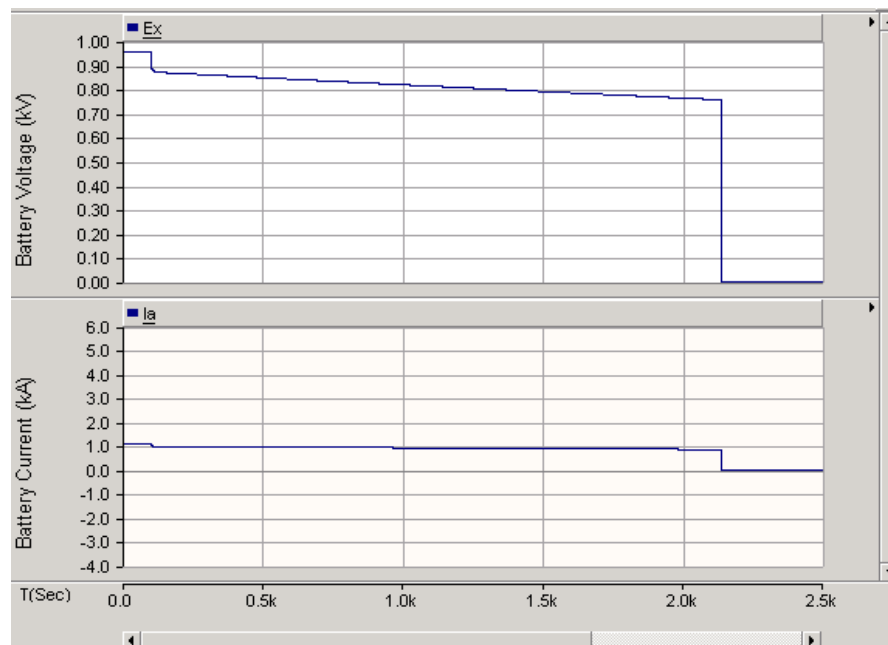


Figure 106. Voltage & Current Waveforms of the 1 MWh Battery

8.5.1.6.3 Transient Analysis of the Single and Stacked Cell Models

The single cell and stacked cell models were analyzed with the resistive load for their steady state performance. The single phase fault was applied to the system in the figure below, near the battery module at 500 seconds.

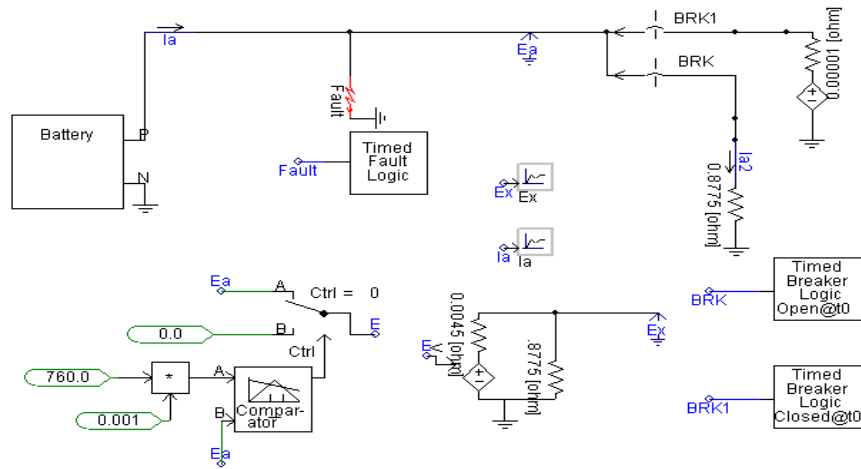


Figure 107. Fault Applied to 1 MWh Battery Equivalent Single Cell Model in PSCAD

The response of the system is plotted below.

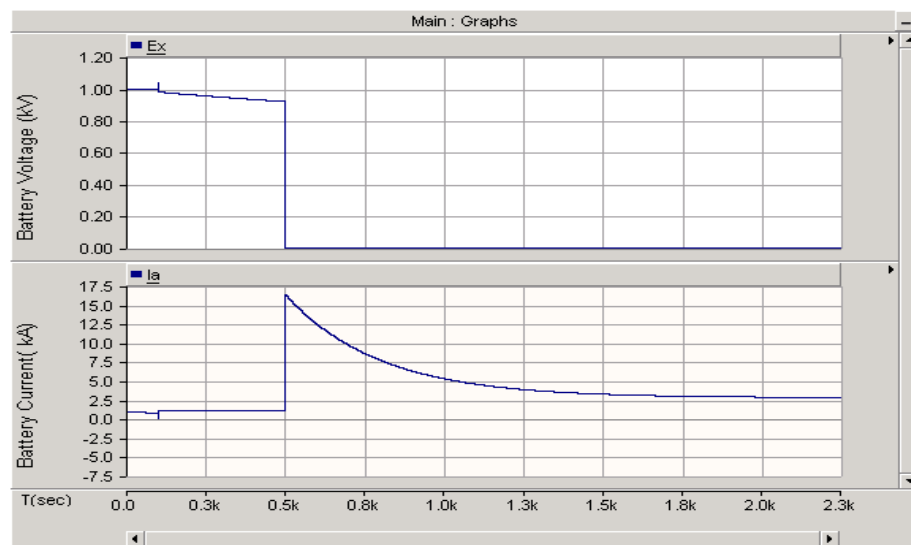


Figure 108. V& I Waveforms of the Single Cell Model

Similarly, the single phase fault was applied near the battery module, at 1500 seconds, to the system as shown below.

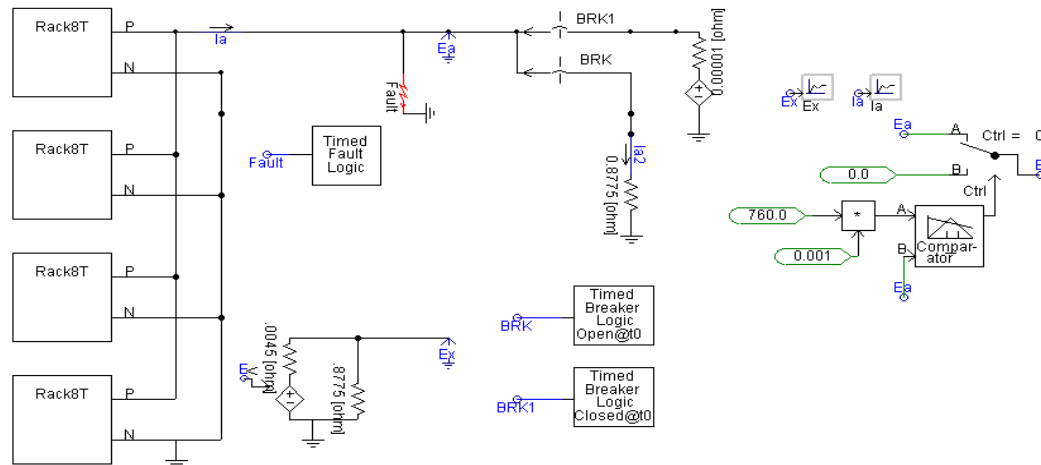


Figure 109. Fault applied to 1 MWh Battery stacked cell Model

The response is shown below and it was observed that both systems were discharging until applying the fault after which the system dropped to zero.

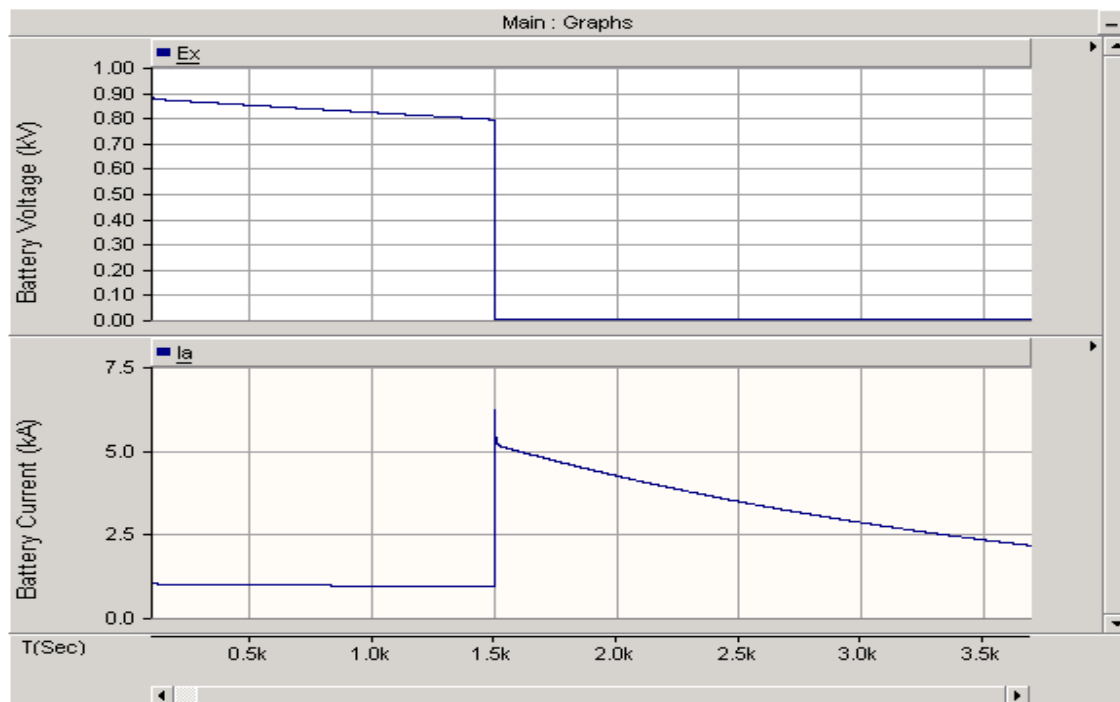


Figure 110. V&I Waveforms of the Stacked Cell Model

The battery models were compared and analyzed on their performance during steady state and transient conditions when connected to the resistive load at 1C discharge rate. It was compared for the performance of the battery models as a single cell and as a stacked cell, when connected to a resistive load. It was observed from the simulation results that the electrical equivalent circuit model of a single

cell was more accurate with voltage and current than the stacked model, and it has discharge characteristics, speed, performance, stability and compatibility. It could be easily integrated in future for higher end applications with the grid. The simulation demonstrated that the stacked model had some loss of voltage due to the stacking of cells, and it has node compatibility in PSCAD.

8.5.1.7 Battery Controllers' Comparison

A comparison study on the operational performance of a battery-connected grid system with bi-directional grid-connected inverter and conventional PQ controller in steady state conditions was performed. The two different methods of control were implemented in PSCAD and the results were compared. Initially the battery was modeled using the combination of runtime based and dual polarization technique (electrical equivalent circuit model). The LMO battery cell was modeled for a rated capacity of 1 MWh. The nominal voltage representing the end of the linear zone of discharge characteristics was considered to be 760 V. The capacity of the battery used in the simulation was 1200 Ah and that parameter did not affect the discharge curve. The maximum voltage when the battery was fully charged was considered to be 1053 V. Some of the parameters that were not taken into consideration were the self – discharge resistance, temperature effects and memory effects. Internal resistance and battery capacity was assumed to be constant and did not vary with amplitude of the current ²⁹.

8.5.1.7.1 PQ Controller

The modeled battery shown below was connected to a three-phase inverter. The converted DC signal from the battery was given to a transformer to step up the voltage to 12.5 kV. The inverter was controlled using a simple PQ controller.

²⁹ Anthony W. Ma. "Modeling and Analysis of a Photovoltaic System with a Distributed Energy Storage System." Thesis. California Polytechnic State University, 2012. May 2012. Web. Aug. 2014

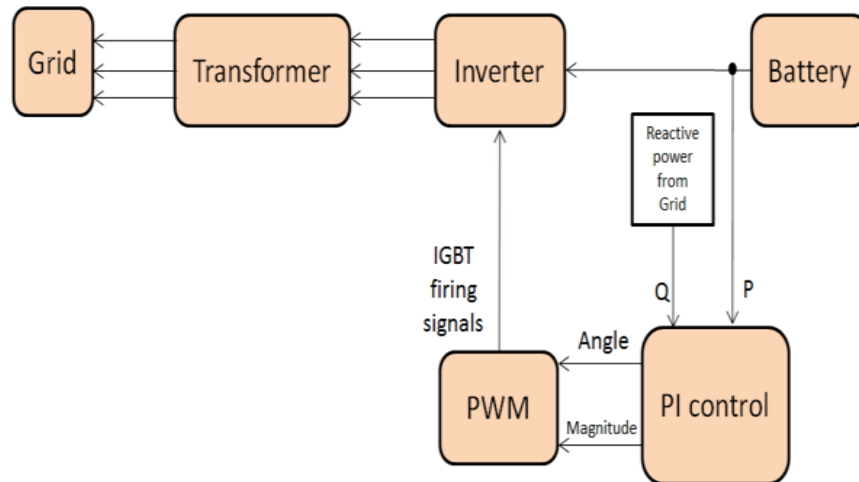


Figure 111. Battery – Grid System with PQ controller

The PI controller block in the above figure was modeled in PSCAD as shown below.

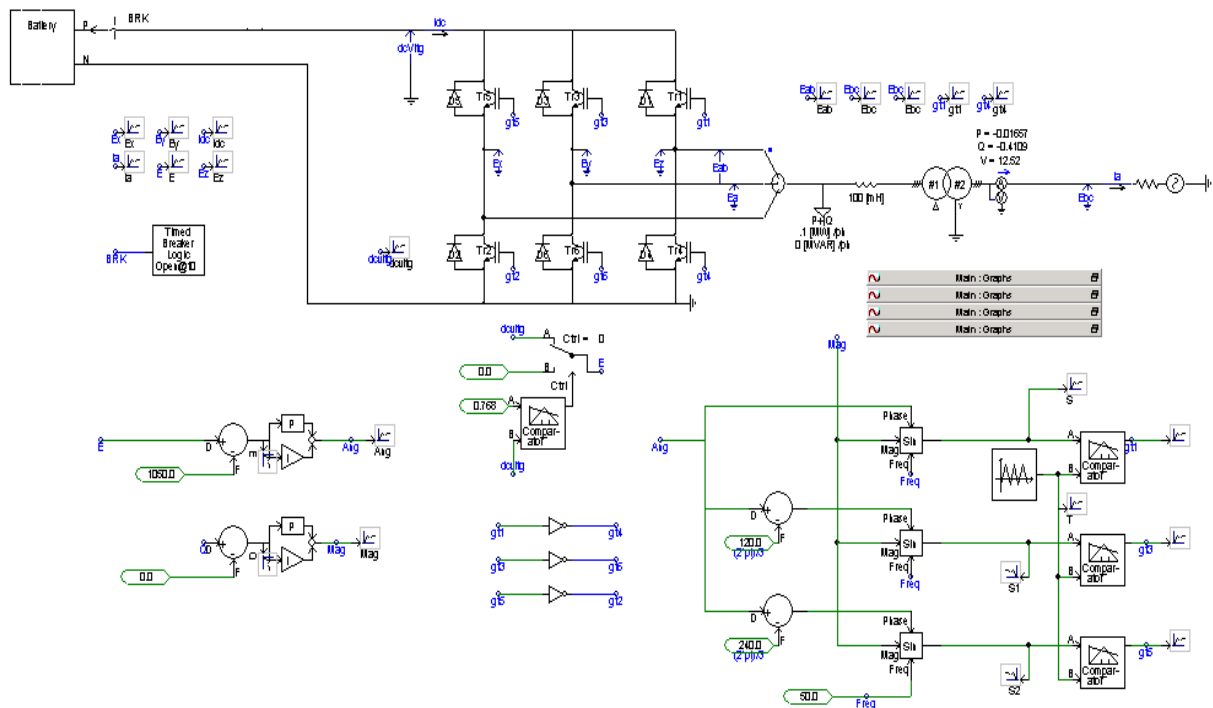


Figure 112. Battery-Connected Grid System using PQ Control in PSCAD

The battery output DC voltage was labeled as “E” and the reactive power was named as “Q”. The measured DC output was compared with the maximum battery voltage and the difference was given to the PI controller as shown below.

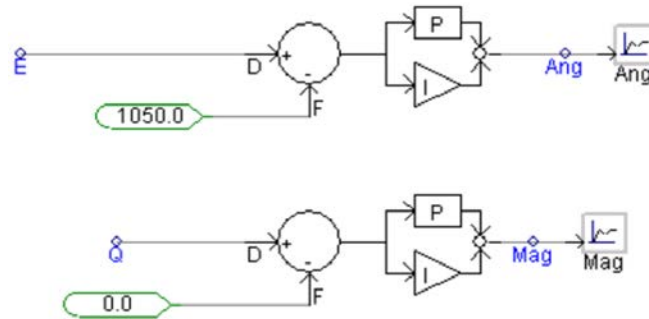


Figure 113. PI controller

The implementation of the controls shows that DC battery voltage, “E”, was found to be 960 V and the grid current, “Ia”, was observed with oscillations as in the figure below.

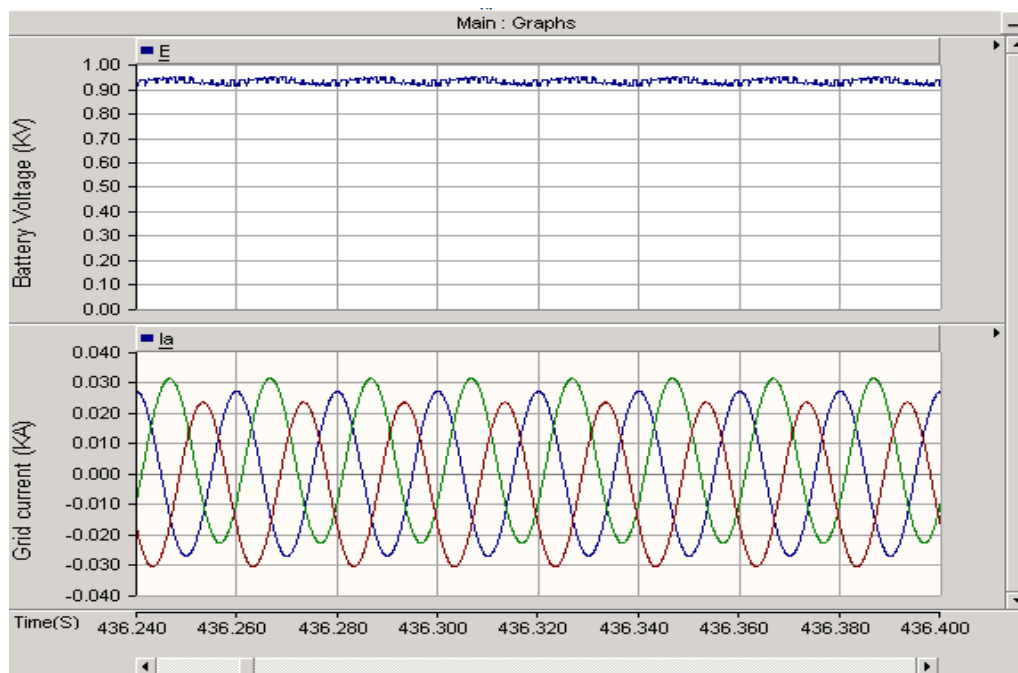


Figure 114. Battery Voltage and Grid Current Waveforms

8.5.1.7.2 Vector Controller

The above system with the PI controller has the buck converter and an inverter which is capable of bi-directional flow to charge/discharge the battery from the grid. The controller was found to be less effective for implementation in the study of advanced battery technology.

In an electric power system the PQ theory is found to be ineffective and hence it has been modified with various referenced current generation techniques. However, with reference current generation, an effective current control technique is required to improve the system performance³⁰. Hence an advanced vector control (current controlled voltage source) was developed and the block diagram was modified as below.

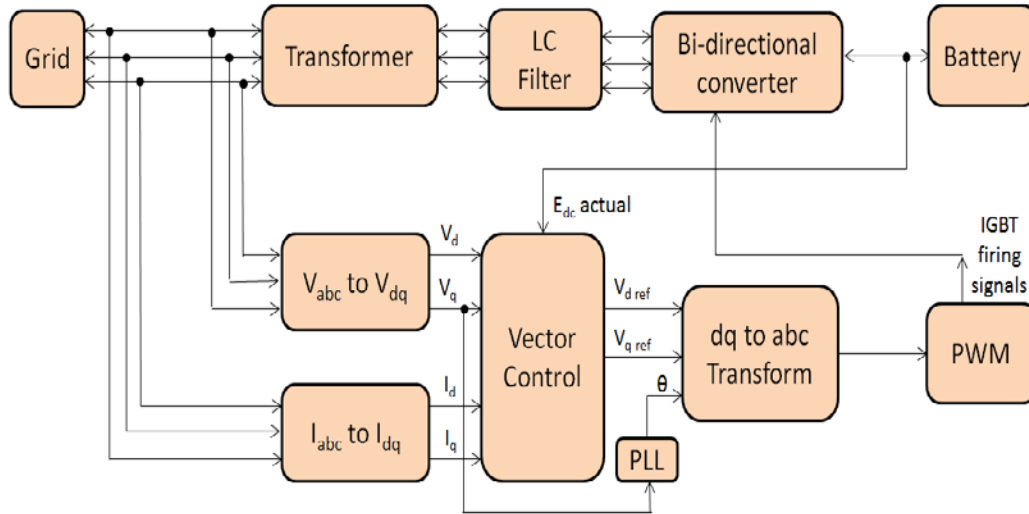


Figure 115. Battery – Grid System with Vector Controller

The battery voltage was given to a bi-directional converter where the DC power initially was converted into DC, and the V_{LLrms} voltage was calculated using the following equation³¹ and given to the LC filter.

$$V_{LL}(rms) = 0.612 m_a v_d$$

Where m_a is amplitude modulation and V_d is battery voltage.

The bidirectional controller was implemented with 6 insulated gate bipolar transistor (IGBT) switches (gt1, gt2, gt3, gt4, gt5, gt6) in parallel with a diode to enable double-sided current flow. The three phase bi-directional converter shown in the figure below had phases controlled with 120-degree shift from each other.

³⁰ Tsengenes, Georgios A., and Georgios A. Adamidis. "22." Performance Evaluation of PI and Fuzzy Controlled Power Electronic Inverters for Power Quality Improvement. INTECH Open Access, 2012. Print.

³¹ Mohan, Ned, Tore M. Undeland, and William P. Robbins. "8." Power Electronics: Converters, Applications, and Design. New York: Wiley, 2002. Print.

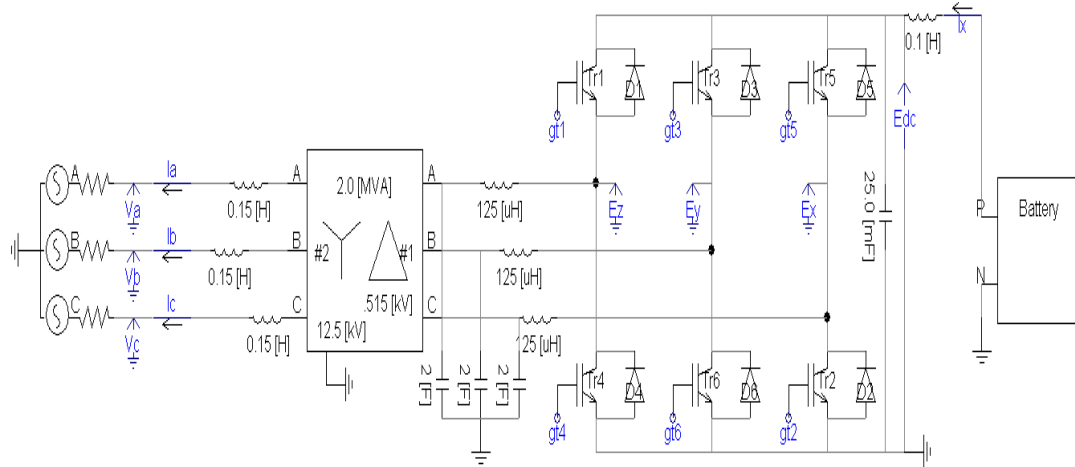


Figure 116. Grid-tied 1 MWh Battery System

The ripple on the total current was relatively small with the small capacitance of 25 mF at the low side for acceptable voltage ripple. The bi-directional converter can operate to have double-sided power flow. The objective of the pulse width-modulated three phase converter was to shape and control the three phase output voltages in magnitude and frequency with a constant input voltage of V_d to be 1050 V. The converter shown above consisted of three legs for each phase. The output was independent of the output load current since one of the two switches in a leg was always 'on' at any instant. In order to produce a sinusoidal output waveform at a desired frequency, a sinusoidal control signal at the desired frequency was compared with the triangular waveform switching frequency and was kept constant along with its amplitude V_{tri} . The triangular waveform was V_{tri} switching at a frequency of 750 Hz to switch the converter switches. The control signal $V_{control}$ was used to modulate the switch duty ratio and had a frequency of 60 Hz which is the desired fundamental frequency of the converter voltage output. Using the amplitude modulation equation, the value was calculated to be 0.8. The frequency modulation m_f was calculated to be 12.5. The voltage at the battery terminal was found to be 1 MW with voltage of 1053 V and 1200 A. The DC power was converted into AC power and the $V_{ll}(rms)$ was calculated to be 514.08 V with the amplitude modulation of 0.8. The voltage was stepped up through a transformer and connected to the grid at 12.5 kV.

The measured values were converted from abc to $\alpha\beta$ transformation³². The obtained transformation was transformed to dq as below where U_d and U_q are the coordinates of the dq rotating reference frame

³² "Documentation Center." Perform Transformation from Three-phase (abc) Signal to $\alpha\beta 0$ Stationary Reference Frame or the Inverse. Mathworks, 13 Aug. 2014.
<http://www.mathworks.com/help/physmod/sps/powersys/ref/abctoalphabetazeroalphabetazerotoabc.html>.

and ' ωt ' is the angular position of the rotating frame³³. The angle was then calculated using the third equation with U_d and U_q values.

$$U_d = \cos(\omega t) U_\alpha + \sin(\omega t) U_\beta$$

$$U_q = -\sin(\omega t) U_\alpha + \cos(\omega t) U_\beta$$

$$\angle U = \text{atan2} \left(\frac{U_q}{U_d} \right)$$

The transformed d and q parameters were fed into the vector control as shown in the figure below.

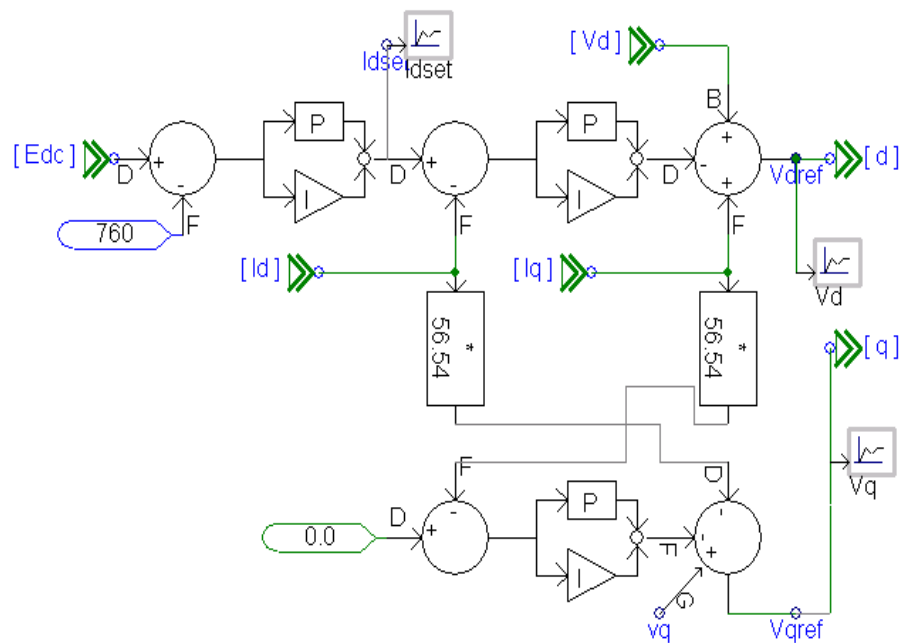


Figure 117. Grid-side Vector Controller

The current reference was obtained from the DC voltage control loop with E_{dc} (battery output voltage). The voltage was manipulated to achieve the reference current in the current control loop. The vector control was built based on the current control³⁴.

³³ "Documentation Center." Perform Transformation from $\alpha\beta 0$ Stationary Reference Frame to Dq0 Rotating Reference Frame or the Inverse. Mathworks,

<<http://www.mathworks.com/help/physmod/sps/powersys/ref/alphabetazerotodq0dq0toalphabetazero.html>>.

³⁴ M, Devadarshanam, Manjunatha B.M, and Damodaram K. "Modeling and Simulation of PWM Line Converter Feeding to Vector Controlled Induction Motor Drive." International Journal of Advanced Research in Computer Science and Software Engineering 2.8 (2012): 321-327.

$$V_d = R i_d + L \frac{di_d}{dt} - \omega L i_q + u_d$$

$$V_q = R i_q + L \frac{di_q}{dt} - \omega L i_d + u_q$$

Where V_d , V_q and i_d , i_q are the voltages and currents of the dq reference frame.

$$\begin{bmatrix} V_{a1} \\ V_{b1} \\ V_{c1} \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \cos \left(\theta - \frac{2\pi}{3} \right) & -\sin \left(\theta - \frac{2\pi}{3} \right) \\ \cos \left(\theta + \frac{2\pi}{3} \right) & -\sin \left(\theta + \frac{2\pi}{3} \right) \end{bmatrix} \begin{bmatrix} V_d \\ V_q \end{bmatrix}$$

The angle “ang, θ ” from the above equations was obtained from the phase locked loop (PLL). Hence the produced angles, V_d and V_q , are converted into V_{abc} values as shown below to get the V_{a1} , V_{b1} and V_{c1} . These voltages were used to switch the IGBT signals of the inverter. The sine wave was compared with the triangular waveform generator switching at higher frequency at 750 Hz with initial phase angle of 0 deg and 50% duty cycle with maximum and minimum output level to be -1 and +1 respectively.

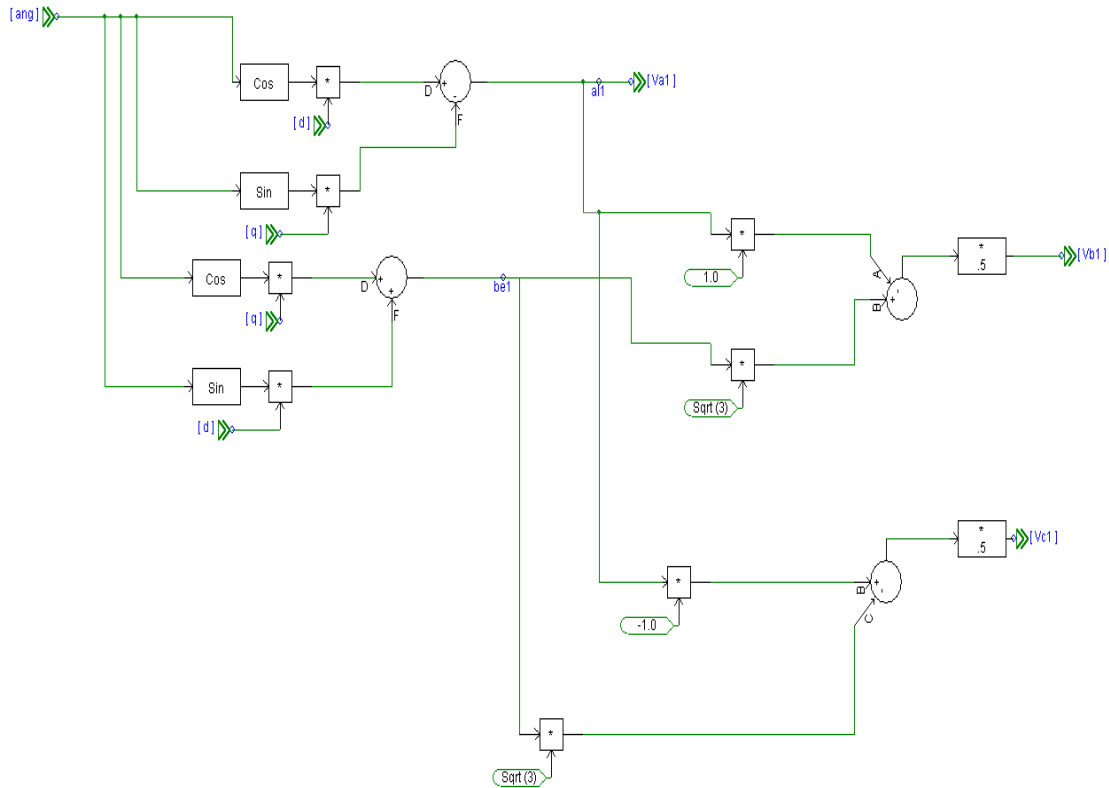


Figure 118. dq-abc Transformation

The signals were used to fire the IGBT switches in the converter to enable the battery charge and discharge performance of the battery. Also it was observed that when the battery was in its discharging state, the vector controller produced better results of grid voltage and current as shown below when compared to that of the PQ controller.

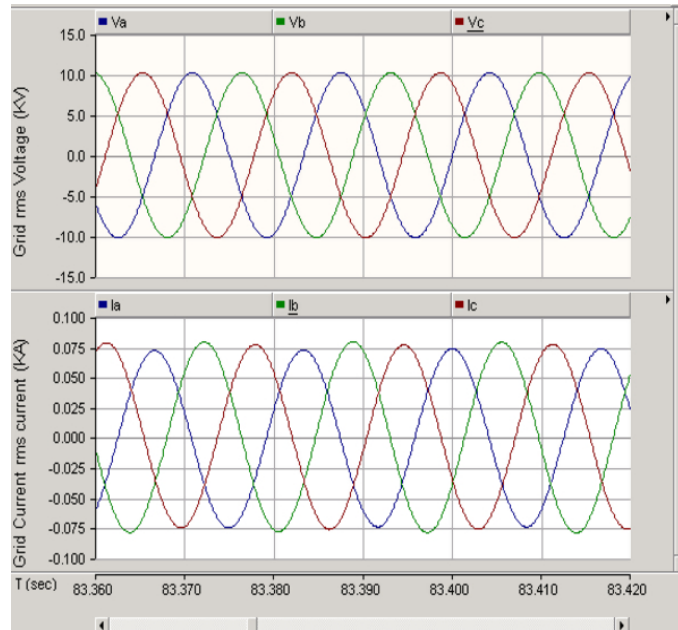


Figure 119. Voltage and Current Waveforms at the Grid

The simulation results of the PQ control and vector control were compared and it was observed that the vector control produced more accurate results than the PQ control. The use of current control and voltage control in the vector control method enables the battery to operate at different charge rates without much change to the system except for the setting of the current reference generated from the voltage control. Thus the bi-directional converter using vector control for the battery to grid application provided greater efficiency, system stability and reduced power oscillations to the system.

8.5.1.8 System Analysis

8.5.1.8.1 Grid Connected Battery System

The vector control method was used to fire the IGBT switches of the converter. The converter acted both as an inverter (DC to AC) and rectifier (AC to DC) to enable bi-directional energy flow between the grid and the battery. With reference to the vector control assembly³⁵, the implication with vector controls was used to simulate the grid-tie battery system for voltage and current control. The voltage

³⁵ NREL, Dynamic Models for Wind Turbines and Wind Power Plants (NREL/SR-5500-52780), October 2011, [Online]. Available: <http://www.nrel.gov/docs/fy12osti/52780.pdf>. [Accessed 21 May 2014]

from the grid was obtained and the V_a , V_b and V_c were transformed into dq values using the following equations ³⁶:

$$V_d = \frac{2}{3} (V_a \sin(\omega t) + V_b \sin\left(\omega t - \frac{2\pi}{3}\right) + V_c \sin\left(\omega t + \frac{2\pi}{3}\right))$$

$$V_q = \frac{2}{3} (V_a \cos(\omega t) + V_b \cos\left(\omega t - \frac{2\pi}{3}\right) + V_c \cos\left(\omega t + \frac{2\pi}{3}\right))$$

$$V_0 = \frac{1}{3} (V_a + V_b + V_c)$$

The current was obtained from the grid and the I_a , I_b and I_c were transferred into I_d and I_q . The PLL as shown below was used for the PSCAD simulation. The PLL was a control system that generated the output signal to be in phase with the input signal. The PLL constituted the phase detector (PD) block, Loop filter (LF) and the voltage controlled oscillator (VCO). The PD block generated the error directly proportional to the phase difference between input and output voltage. The loop filter reduced the input phase error with AC components. The error produced was given as, $e = V \sin(\theta - \theta')$. The PLL locked the $\theta = \theta'$ such that error=0. The VCO generated the frequency shift and phase outputs with respect to given central frequency as a function of input voltage.

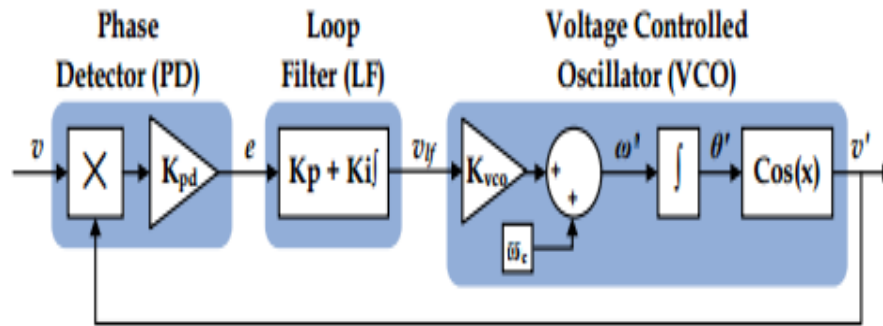


Figure 120. Phase Locked Loop ³⁷

The above simplified PLL block diagram was implemented in PSCAD. The grid voltage V_{abc} transformed into V_q voltage and was multiplied with the generated voltage at the voltage-controlled oscillator (VCO). The error produced at the phase detector was taken as $e(t)$. The proportional (K_p) and integral (K_i) terms of the loop filter are given by the following equation.

³⁶ MathWorks Documentation Center. Perform Park Transformation from Three-phase (abc) Reference Frame to Dq0 Reference Frame. [Online]. Available: http://www.mathworks.com/help/physmod/sps/powersys/ref/abc_to_dq0transformation.html. [Accessed 27 May 2014].

³⁷ Gao, S., & Barnes, M. (2012). Phase-locked loop for AC systems: Analyses and comparisons.

$$e(t) = k_p e(t) + k_i \int_0^t e(t) dt$$

The K_p and K_i were the tuning parameters adjusted to obtain the desired current “ $e(t)$ ” as output. The generated angle was used to convert the dq parameters to abc parameters which were used to switch the IGBTs of the converter.

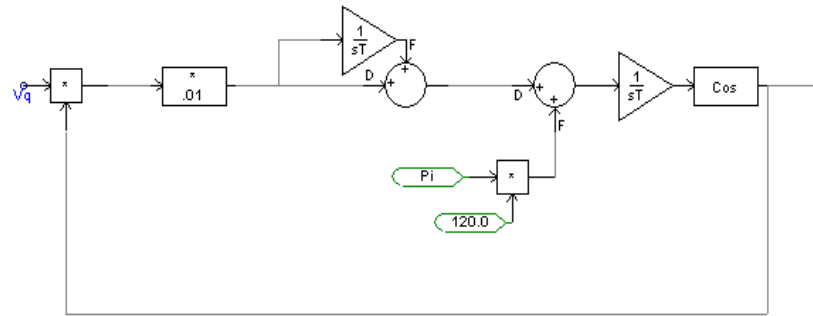


Figure 121. PLL in PSCAD

The PLL was used to control the angle between the inverter and the grid and to considerably change the angle between grid side and inverter side voltage to enable the battery charge and discharge characteristics. The ω_c value was calculated as $2\pi \cdot 60$.

Simulating the system, the battery was charged from 850 V to 1150 V at 3600 seconds at 1C as shown by battery voltage, Edc, and at current of 1200 Ah as shown by battery current, Ix, in the figure below.

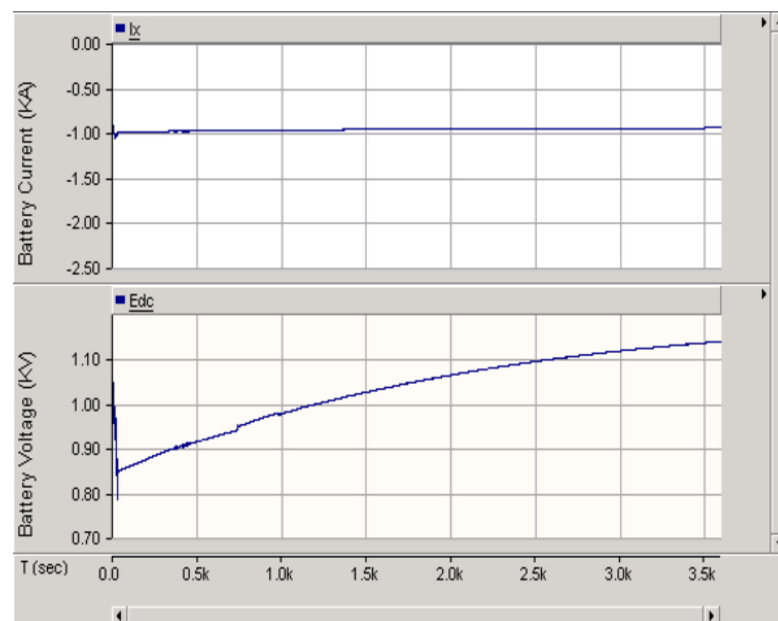


Figure 122. Battery Charge Characteristics at 1C Rate

Similarly the battery discharge was observed from 1050 V to 760 V at 3600 seconds at 1C rate as shown by battery voltage, Edc, and at current of 1200 Ah as shown by battery current, Ix, below.

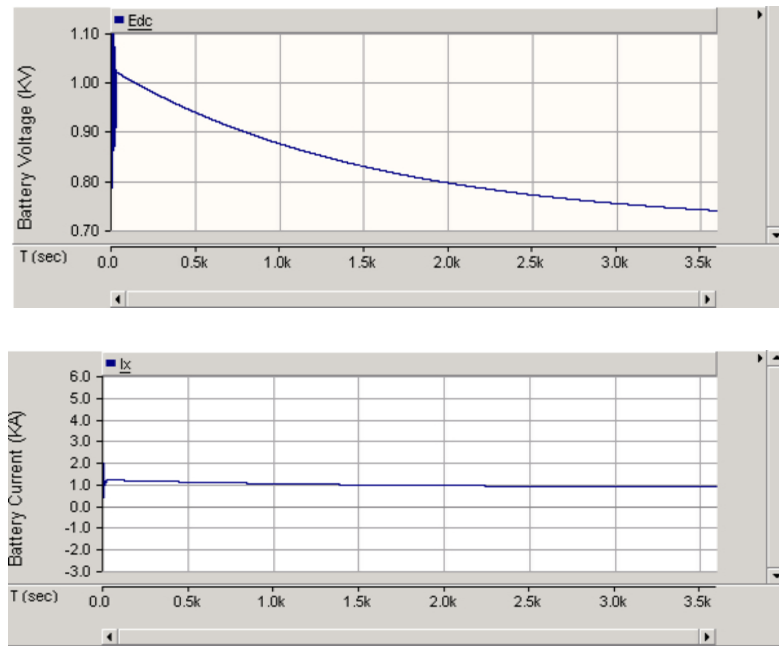


Figure 123. Battery Discharge Characteristics at 1C Rate

8.5.1.8.2 Grid-Connected Wind System

A wind turbine (synchronous generator) was modeled for 1.7 MW power produced with a constant mean wind speed (13m/s) and was connected to the grid at 12.5 kV. The damp wind fluctuations were ignored. The turbine voltage was 690 V and the $V_{DC \text{ bus}}$ was calculated to be 1600 V. With the aim to build the DC bus, the storage capacitance value was calculated to be 1.32 F. The resistor was calculated as 1.948 ohm. The breaker circuit was calculated based on the load time constant, $T_r = 3\tau = 3RC$, and was obtained as 7.71408. An inductor was added in order to model a current source at the input of the inverter. I_{dc} was calculated to be 1062.5 A and the inductance was calculated as 3.014 H³⁸.

³⁸ "PSCAD ® Version 4.2 Wind Turbine Applications Technical Paper". PSCAD: Power System Simulation. Cedrat, Manitoba HVDC Research Center Inc., Jan. 2006. Web. 17 June 2014.

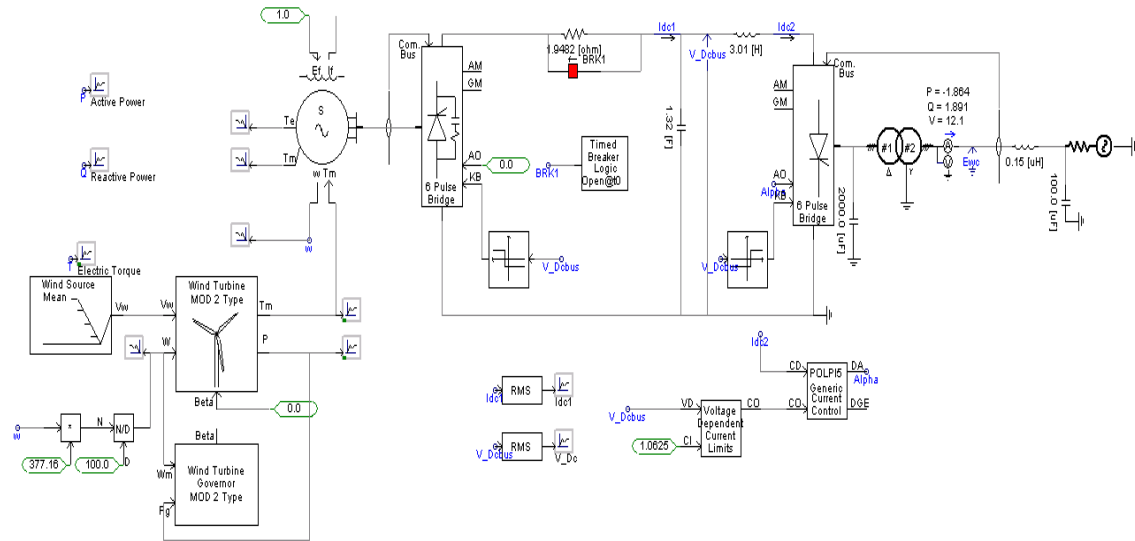


Figure 124. Wind Grid-Tied System in PSCAD

The machine rated angular mechanical speed was calculated to be 3.77 rad/s. The number of pole pairs used was 100. The rotor radius and rotor area were 31.77 m and 3171.2 m² respectively. The air density was considered to be 1.225 kg/m³ with the gear ratio as 1. The wind turbine governor component shown in the above figure had the variable pitch control enabled with rated electrical efficiency frequency of the machine as 3.7716 rad/s. The synchronous generator used in the simulation was rated for voltage of 0.69 kV and current at 821.25 A. The proportional gain and integral gain values of the PI regulator parameters were 6.2 degree/pu. The speed damping parameters were set to zero. The blade actuator parameters had the pitch angle lower limit and pitch angle upper limit set to be 0 and 25 degrees, respectively. The figure above shows the wind turbine model configuration where the electrical power created by the synchronous generator is diode rectified to DC, switched into three-phase AC with a thyristor inverter and connected to the grid via a step-up transformer. The figure below presents the performance of the above model operating at 1.7 MW capacity under steady state wind operation. The parameters such as turbine power (1.7 MW), active power (1.5 MW), reactive power (≈ 0 MW), electric torque (490 k NM) and mechanical speed (3.4 rad/sec) were observed from simulation.

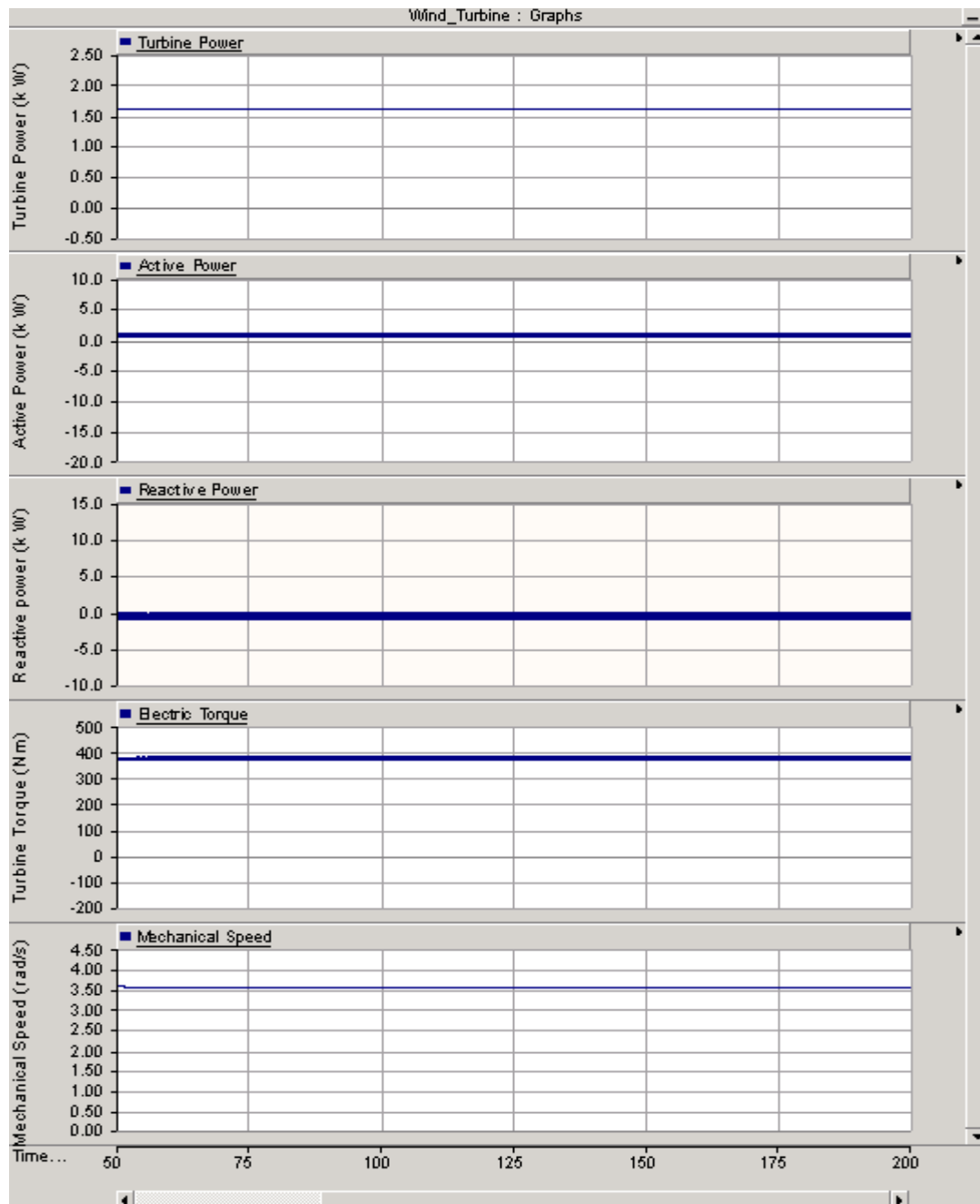


Figure 125. Wind Turbine Model Parameters

The rms grid voltage (V_{g_rms}) was 12.5 kV. The V_{base} was calculated as 10.2 kV. The simulation results observed at the grid are shown below. The current obtained when simulating the wind turbine connected grid system was 580 A.

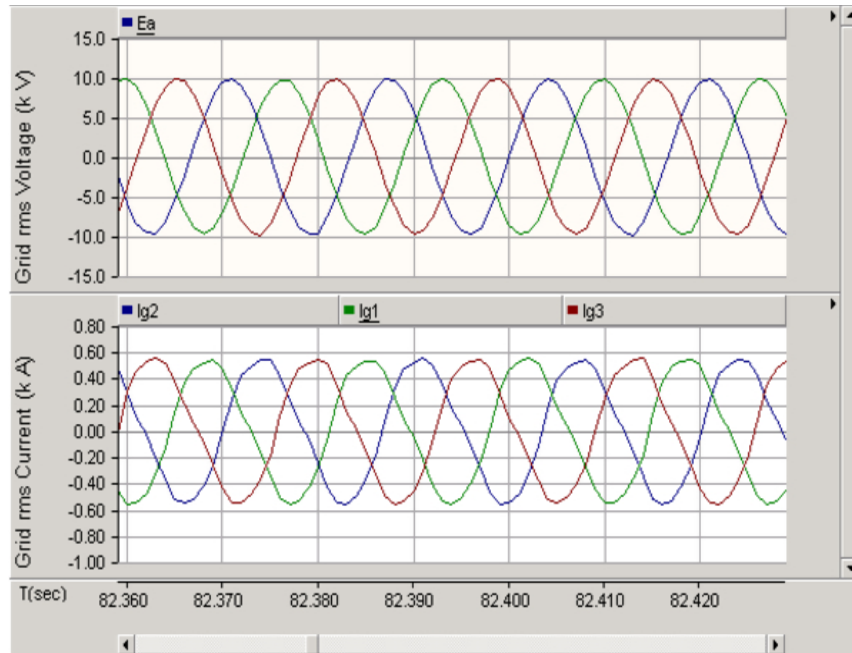


Figure 126. Voltage and Current at Grid

8.5.1.8.3 Grid Connected Battery System – Fault Analysis

The battery system was connected to the 12.5 kV grid through a resistor of 0.1 ohm. A three-phase fault to ground was applied near the grid to the grid-connected battery system as shown below.

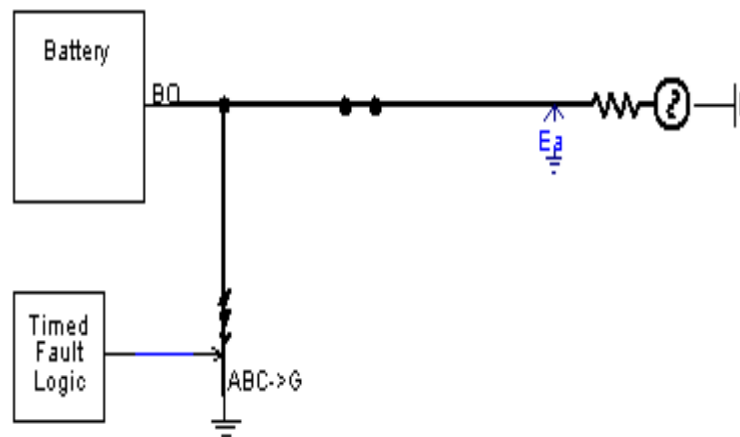


Figure 127. Battery to Grid – Fault

The transient state analysis of the battery discharge characteristic shown below was observed to be ≈ 1000 V until time at 110 seconds and had a very small reduction in voltage level during the three-phase fault period applied between 110-113 seconds. The voltage resumed back after the fault period and

attained its steady state condition. The battery current was found to be increasing during the fault period and settled after the fault at 1180 A.

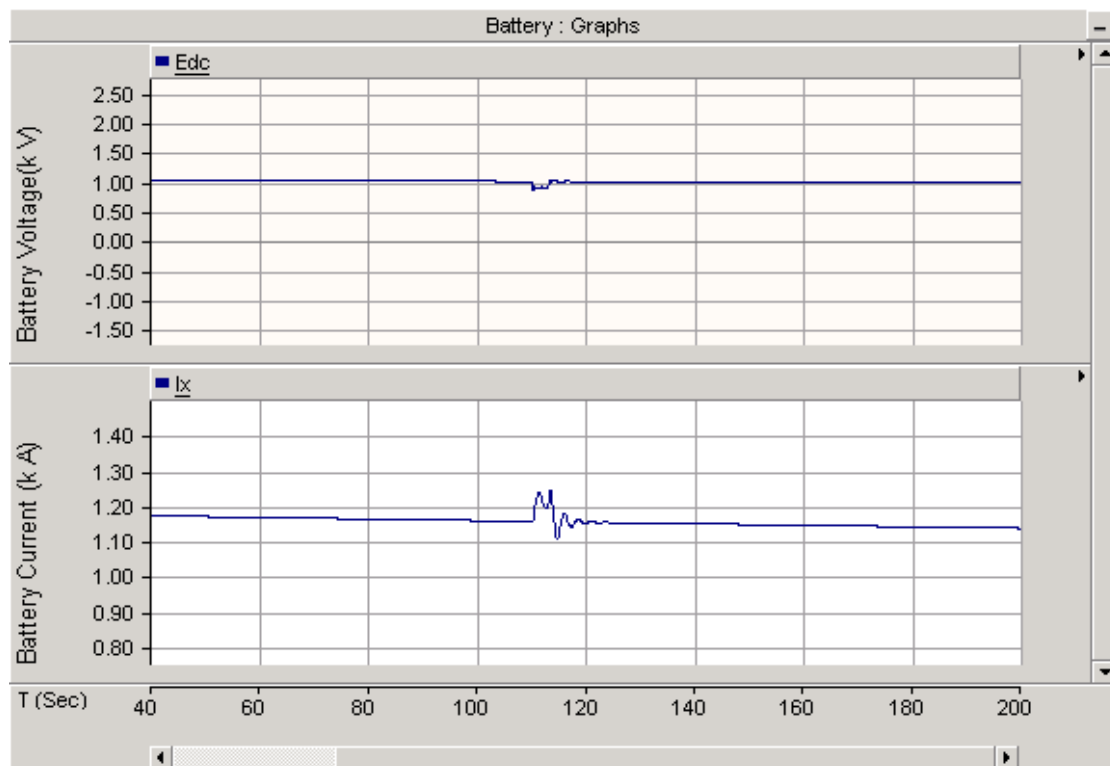


Figure 128. Battery Response— Fault

The plots in the next figure show the grid response while the battery system was connected to it during transient conditions. The grid rms voltage was found to be 12.2 kV and decreased during the fault while the grid rms current increased during the fault.



Figure 129. Grid Response – Fault

8.5.1.8.4 Grid Connected Wind System – Fault Analysis

A three-phase fault to ground is applied between 110 -113 seconds near the grid to the system shown below.

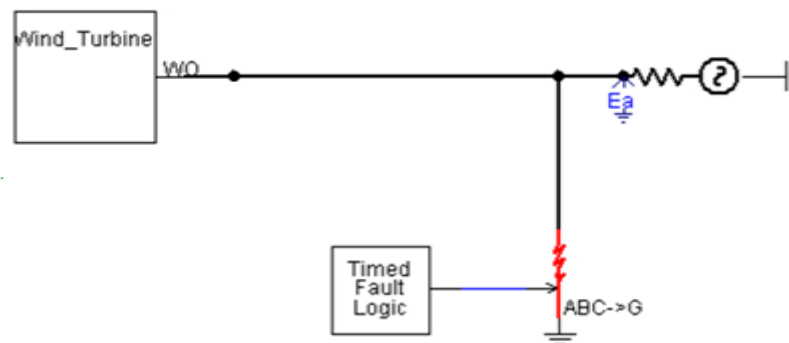


Figure 130. Wind to Grid – Fault

The transient state analysis of the wind connected to the grid shows in the figure below that the turbine power dropped, active power of the generator dropped, mechanical speed of the turbine increased, reactive power of the generator increased, and electric torque of the generator decreased.

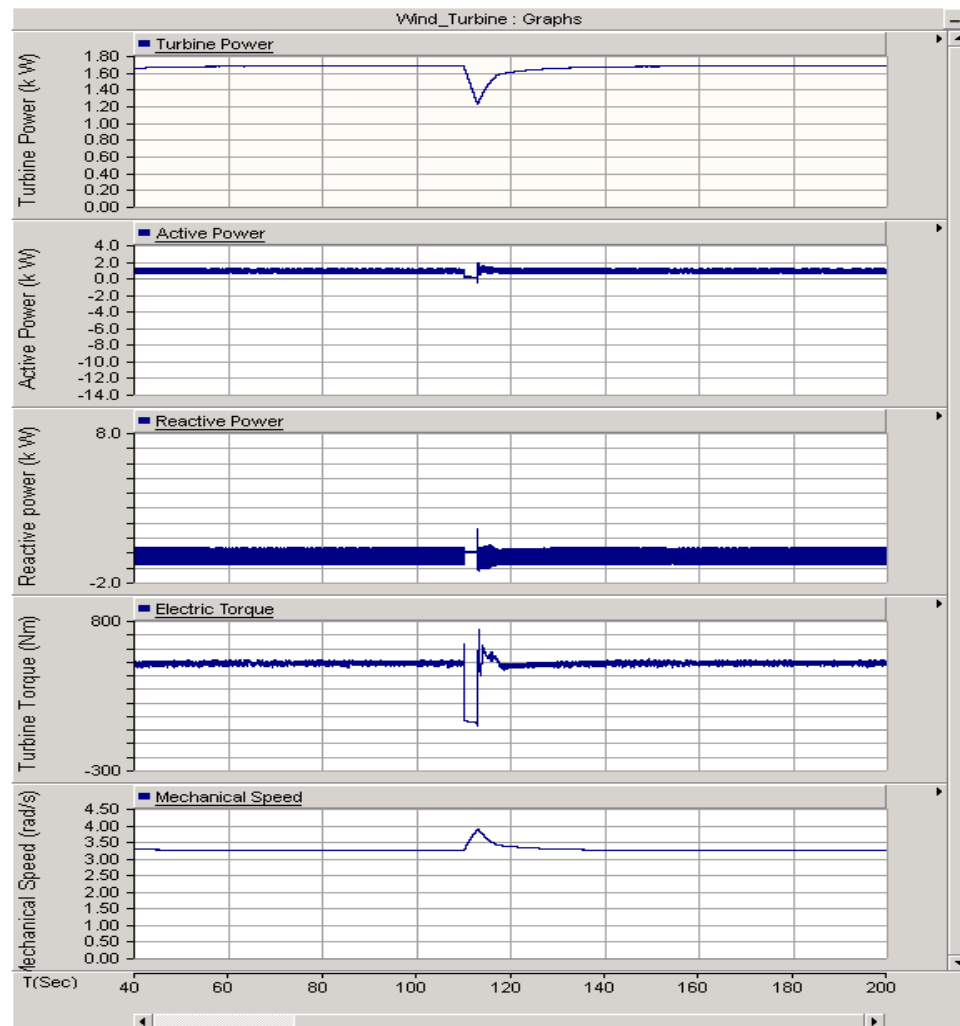


Figure 131. Wind turbine Response - Fault

8.5.1.8.5 Grid Connected Wind System – Fault Analysis

The system shown below was constructed as modules and built with PSCAD using battery and wind modules. The three-phase voltage source model in PSCAD was used as a grid.

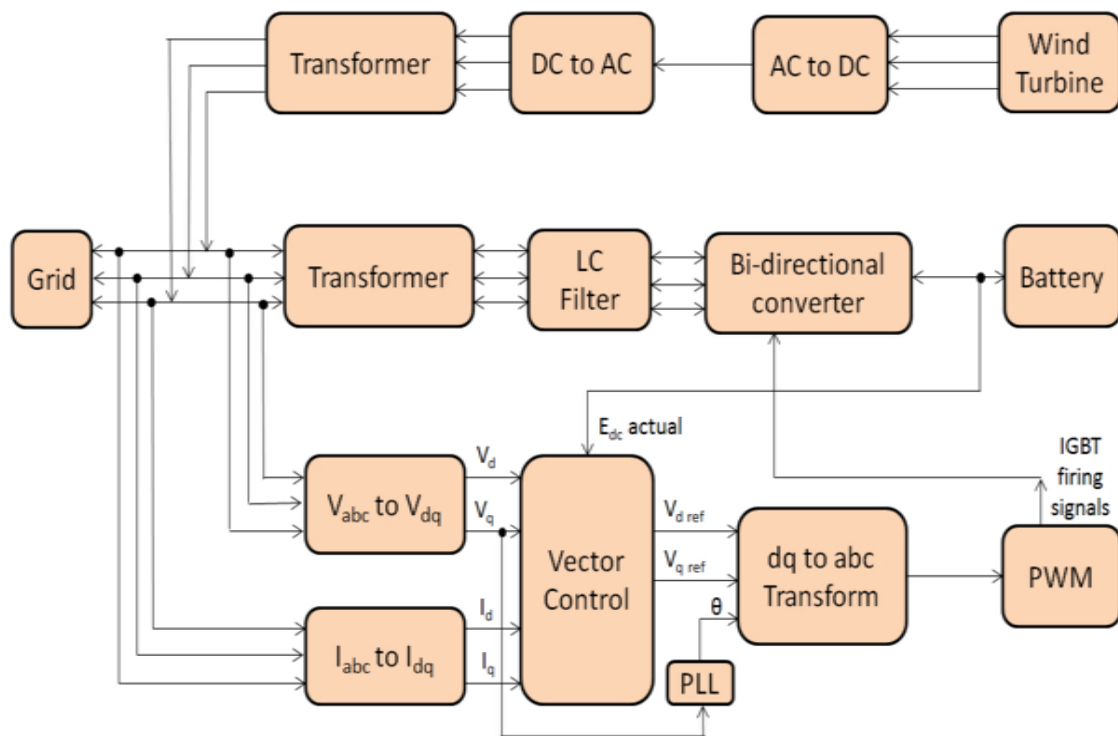


Figure 132. Configuration of Grid-Connected Wind and Battery System

The transient analysis was conducted by simulating the above system and injecting the three-phase faults to determine how the components would perform.

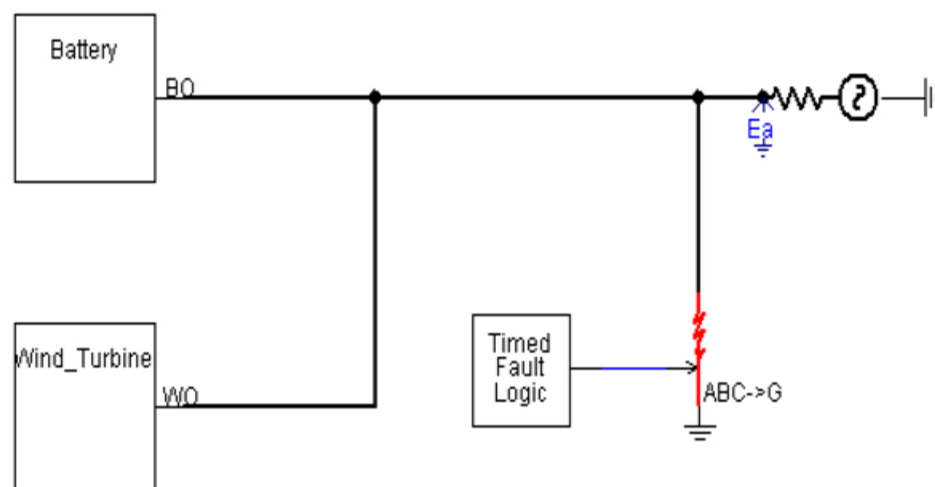


Figure 133. System under Fault Application

A three-phase fault was introduced into the system for a 3-second interval beginning at the 110 second time and the corresponding effects on battery and wind module were observed as shown below; basically, the battery current increased and the battery voltage dropped and had oscillations during the fault.

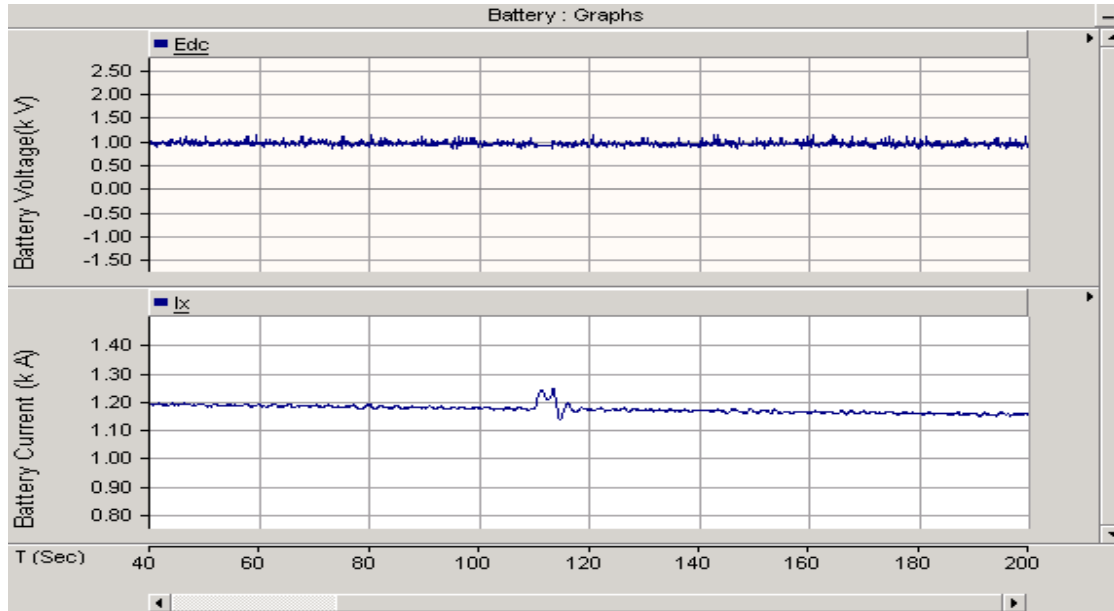


Figure 134. Effects on the Battery

Also it was observed that the turbine power dropped, active power of the generator dropped, mechanical speed of the turbine increased, reactive power of the generator increased, and the electric torque of the generator decreased. These results are evidenced below.

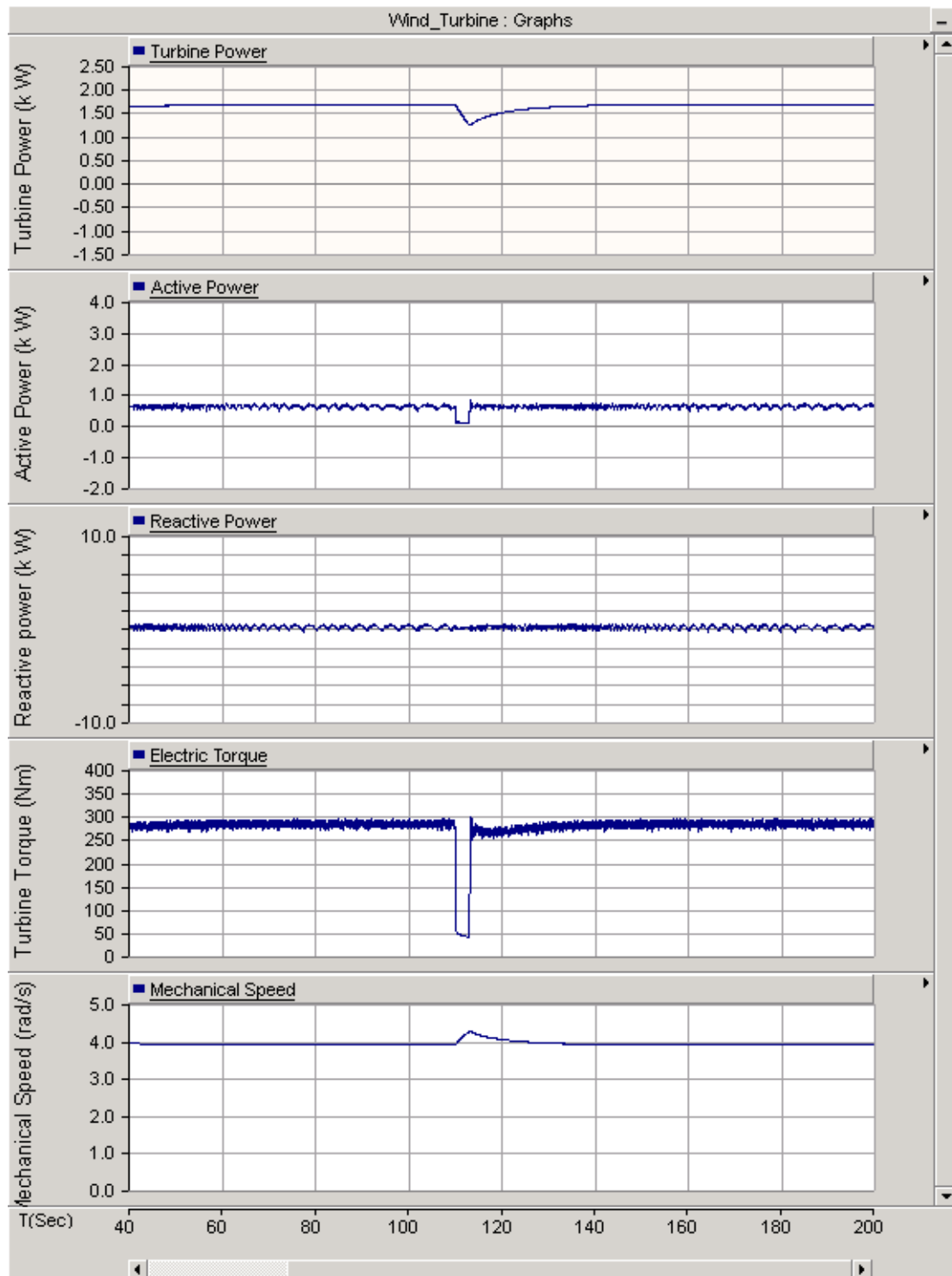


Figure 135. Effects on 1.67 MW Wind Turbine

The three-phase fault introduced into the system for a 3-second interval beginning at the 110 second time and the corresponding effects on the grid when the battery is connected to the system were observed as shown below. The grid voltage dropped and the grid current increased during the fault period.

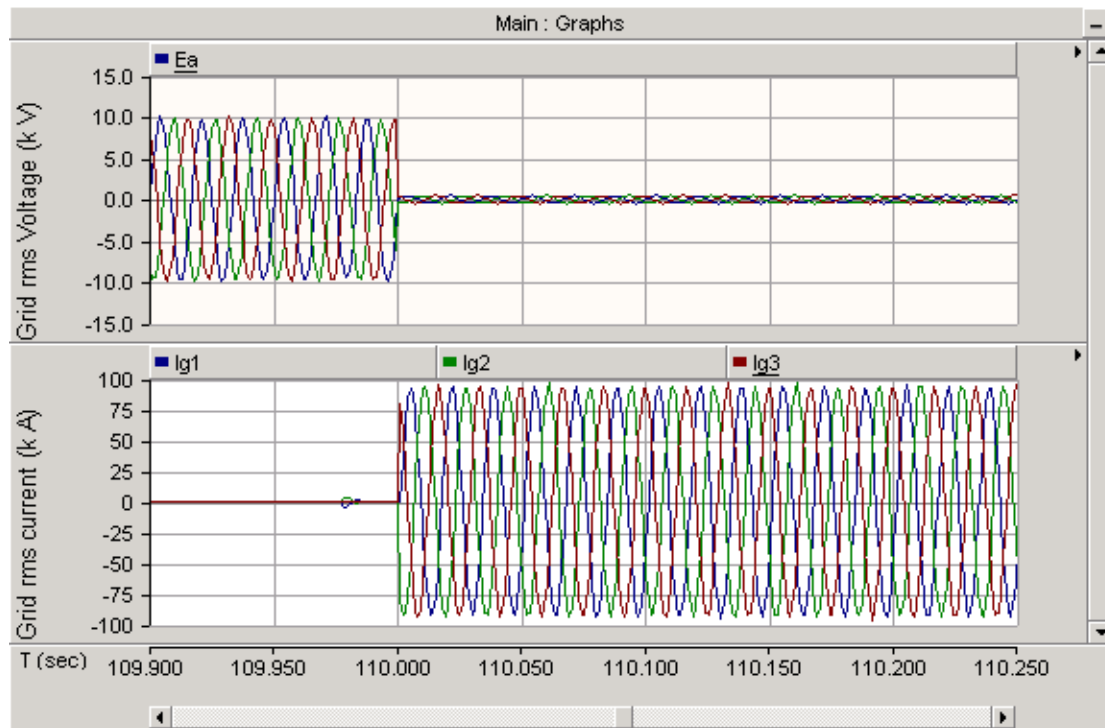


Figure 136. Grid Voltage and Current Response to Fault at 110 Seconds

8.5.1.9 Weak/Strong Grid Analysis

Weak grid is when the voltage and frequency fluctuate, potentially making the system unstable. The strong grid is stable with not much variation in voltage or frequency. If the impedance in the figure below is larger, then fault level and short circuit ratios will be less, and the grid will be considered weaker. The appropriate size of the wind turbine will be such that the value of the short circuit ratio must be around 20.

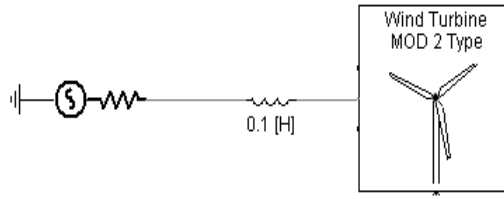


Figure 137. Impedance between Grid and Wind Turbine

If the value of the short circuit ratio is equal to or above 20, then the system is considered strong whereas the value of the short circuit ratio below 20 represents a weak grid. Equations below were used to determine a strong/weak grid:

$$S = \frac{U^2}{Z}$$

Where,

S - Three phase short circuit apparent Power of the grid/fault level

U - r.m.s value of the nominal voltage of the grid

Z - Impedance between source and load ($\sqrt{R^2 + X_L^2}$)

$$\tan(\Psi_k) = \frac{2\pi f_g L}{R}$$

Where,

Ψ_k - network impedance angle/net angle

R, L - selected to obtain the appropriate Ψ_k

f_g - nominal grid frequency (60Hz) ³⁹

The battery storage system constitutes the power of 1 MWh and the wind turbine constitutes the power of 1.7 MW. Thus the system size, S_n , was taken to be the sum of the total power of both battery and wind turbine and was calculated to be 2.7 MW. The grid voltage, U, is 12.5 kV. The net angle is taken as 30 for the calculation. Assuming the short circuit ratio varied from 2 to 30, the impedance values were calculated. Plugging the known values in the above equations, the R and L values were obtained and tabulated as in the figure below.

³⁹ JawwadZafar, Salman Ahmed Rajput, "Integration of Wind Energy Converters into an Existing Distribution Grid, Maximum of Wind Energy Converters for a 40 kV Substation" Chalmers University of Technology, 2005

S _n , System size=2.6MW				
U _i , Grid =12.5KV				
		Net angle		
Short circuit ratio (SCR)	30	30	50	50
	R(ohm)	L(H)	R(ohm)	L(H)
2	26.0224	0.0399	19.3145	0.0611
4	13.0112	0.0199	9.6573	0.0305
6	8.6741	0.0133	6.4382	0.0204
8	6.5056	0.01	4.6286	0.0153
10	5.2045	0.008	3.8629	0.0122
12	4.3371	0.0066	3.2191	0.0102
14	3.7175	0.0057	2.7592	0.0087
16	3.2528	0.005	2.4143	0.0076
18	2.8914	0.0044	2.1461	0.0068
20	2.6022	0.004	1.9315	0.0061
22	2.3657	0.0036	1.7559	0.0056
24	2.1685	0.0033	1.6095	0.0051
26	2.0017	0.0031	1.4857	0.0047
28	1.8587	0.0028	1.3796	0.0044
30	1.7348	0.0027	1.2876	0.0041

Strong grid

Figure 138. Impedance Table for Different SCR Values

The resistor and inductor values were plotted for varying SCR values. The values from the above figure were plotted as a graph shown below and it was observed that the lower the R and L values the stronger the grid system with high SCR values more than 20.

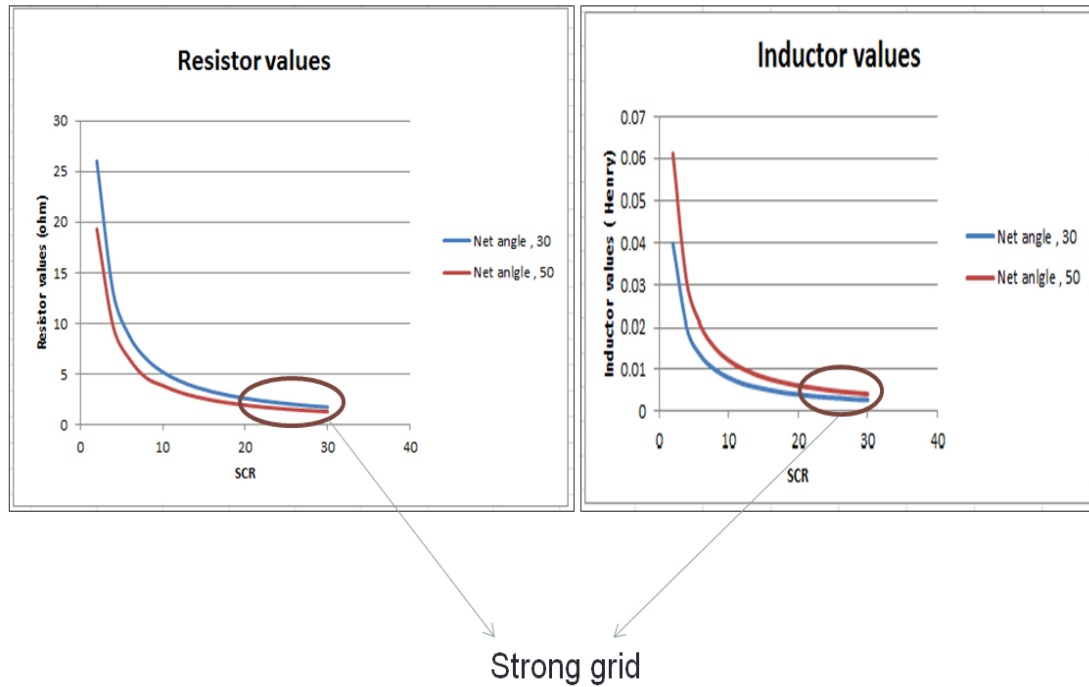


Figure 139. Graph on Strong/Weak Grid

8.5.1.9.1 Grid Connected to Battery System

The battery system was connected to the 12.5 kV grid through a variable inductor and a variable resistor. The multiport switch modeled in PSCAD as shown below enabled the variance of R and L values according to the SCR range of 8 to 26. Thus the project settings in PSCAD were adjusted to have multiple runs so that for each run the SCR value changed and thereby took the corresponding R and L values through the multiport switch. The battery system was initially connected to a weak grid for the first run with an SCR value of 8 and the corresponding calculated R and L values as 6.5056 ohm and 0.01 H, respectively. The use of a multiport switch enabled the change in the impedance values and caused the battery system to be connected to a strong grid.

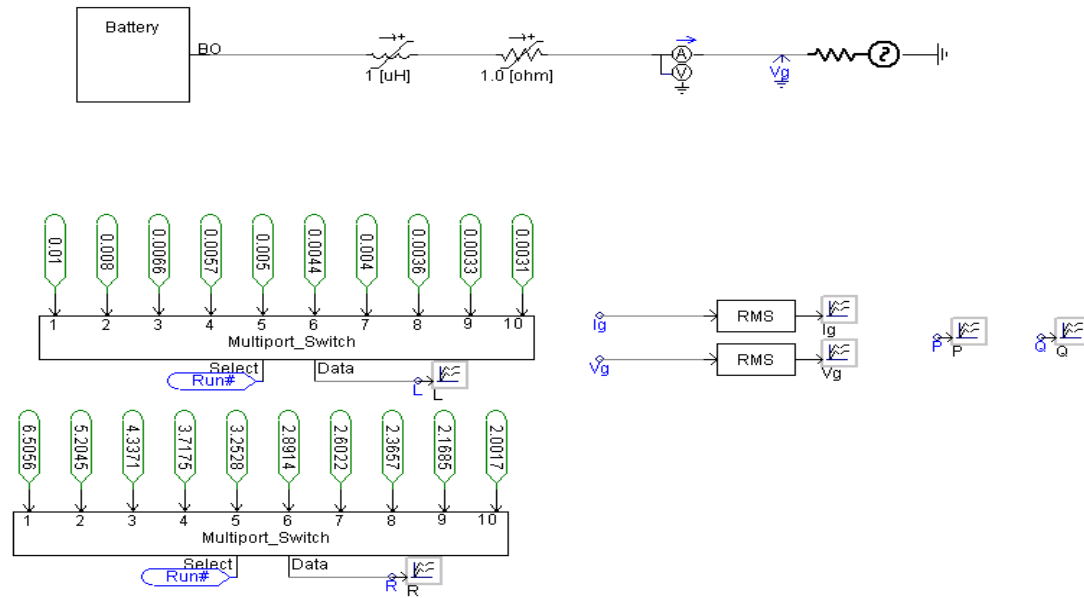


Figure 140. Battery to Grid

The response of the battery and grid parameters were observed. The multiple run values were saved to a file and the data was extracted through MATLAB and plotted as shown below. The steady state analysis of the battery connected to grid did not show any variation with the values changed from weak to strong grid.

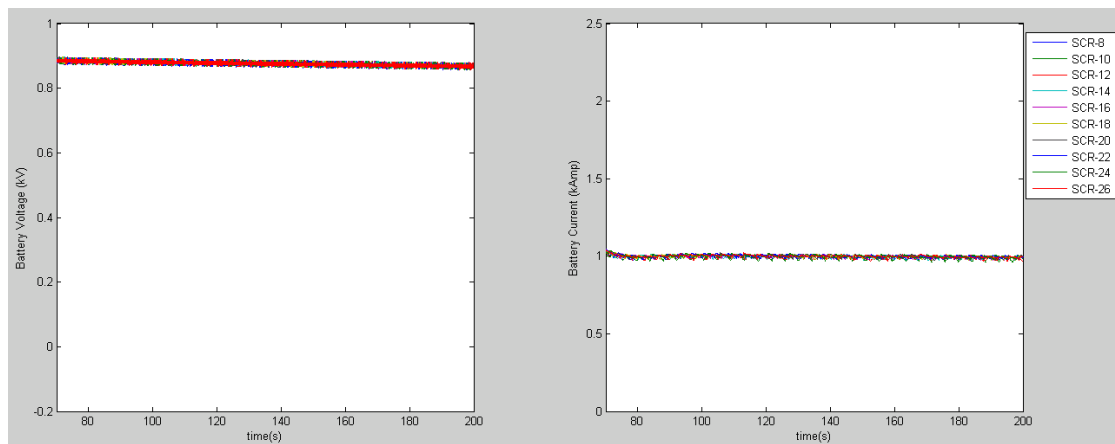


Figure 141. Battery Voltage and Current

The battery discharge was observed in the battery voltage in the left figure and was observed to be 960 V until 200 seconds and continued to discharge until the nominal voltage at 3600 seconds. The battery current in the right figure was found to be settling at 1100 A.

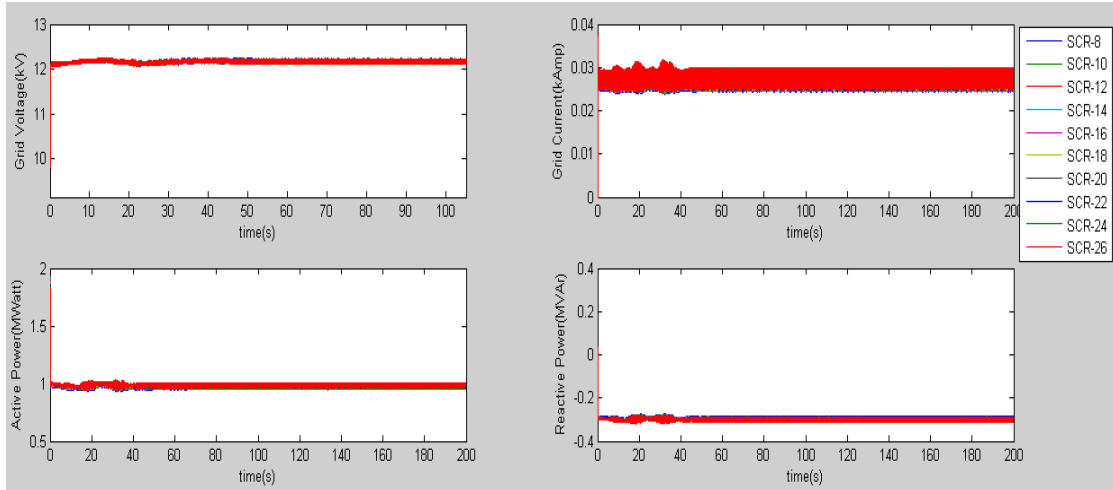


Figure 142. Grid - Voltage and Current (top) & Active and Reactive Power (bottom)

The plots above shows the grid response when a battery system is connected to it. The grid rms voltage was found to be 12.2 kV and the grid rms current was 30 A. The active power at the grid was 1 MW and the reactive power at the grid was -0.3Mvar. The response graphs in the above figures show the legend with the SCR varying from 8 to 26. The legend was used to observe the response of the system when it transferred from a weak to strong grid. Both the grid and battery response did not show much variation during that transition from weak to grid.

8.5.1.9.2 Grid Connected to Battery System-Fault

A three phase fault to ground was applied near the grid to the grid connected battery system. The transient state analysis of the battery connected to grid showed variations when the impedance values were changed from weak to strong grid through the SCR. The plots below show the grid response when the battery system was connected during transient conditions. The grid rms voltage was found to be 12.2 kV and decreased during the fault. The grid rms current was 30A initially but increased during the fault. The active power at the grid reduced and the reactive power at the grid increased during the fault and settled after oscillations. Both the grid and battery discharge characteristics showed considerable changes in transition from weak to strong grid during the time when the fault was applied. The figure below shows the zoomed-in view of the grid voltage and current for a weak grid with SCR taken to be 8 (weak grid). The active and reactive power was found have high oscillations during the fault.

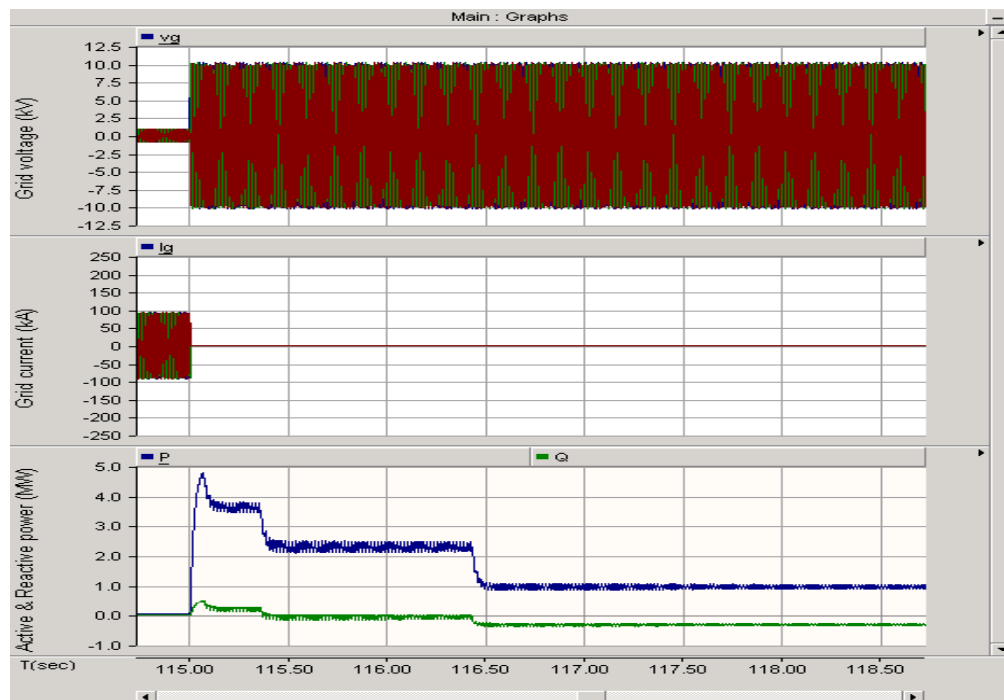


Figure 143. Response at the Grid during Fault (Weak)

The following figure is simulated for SCR to be 26 (strong grid). Thus it is evident that the strong grid sustained the oscillations and tended to reach steady state faster as compared to a weak grid. The oscillations were found to be comparatively less in a weak grid.



Figure 144. Response at the Grid during Fault (Strong)

The battery discharge characteristic in the following figure was simulated for weak grid and was observed to be 1000 V until time at 110 seconds and had a reduction in voltage level during the three-phase fault period applied between 110-113 seconds. The voltage resumed back after the fault period and attained its steady state condition. The battery current increased to 1180 A during the fault period and settled after the fault at 1100 A.

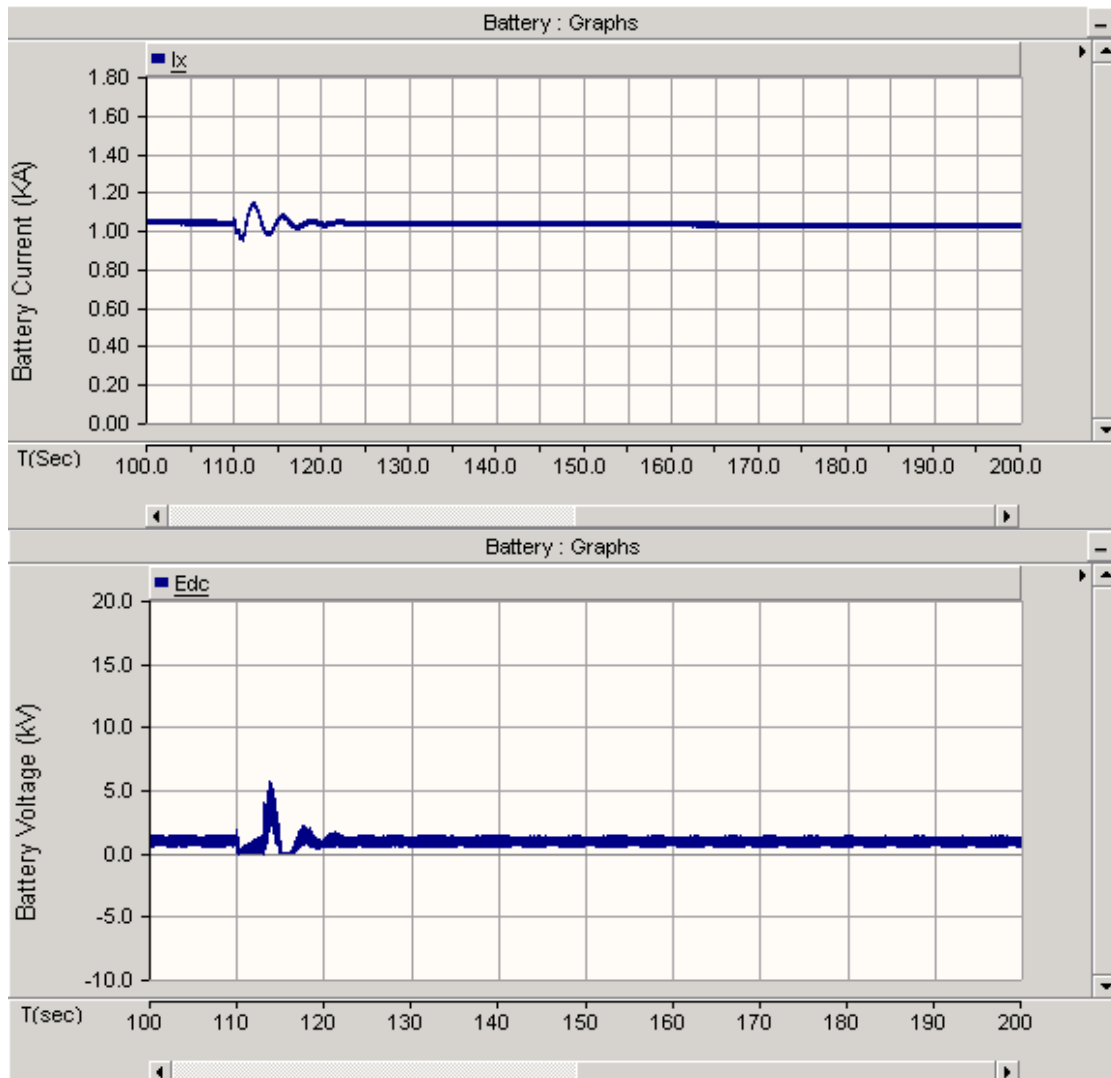


Figure 145. Response at the Battery during Fault (Weak)

The battery discharge characteristic in the following figure was simulated for strong grid and is observed to be 1000 V until time at 110 seconds and had very small distortions and oscillations in voltage level during the three-phase fault period applied between 110-113 seconds. The voltage resumed back after the fault period and attained its steady state condition. The battery current was increased to 1300 A during the fault period and settled after the fault at 1100 A.

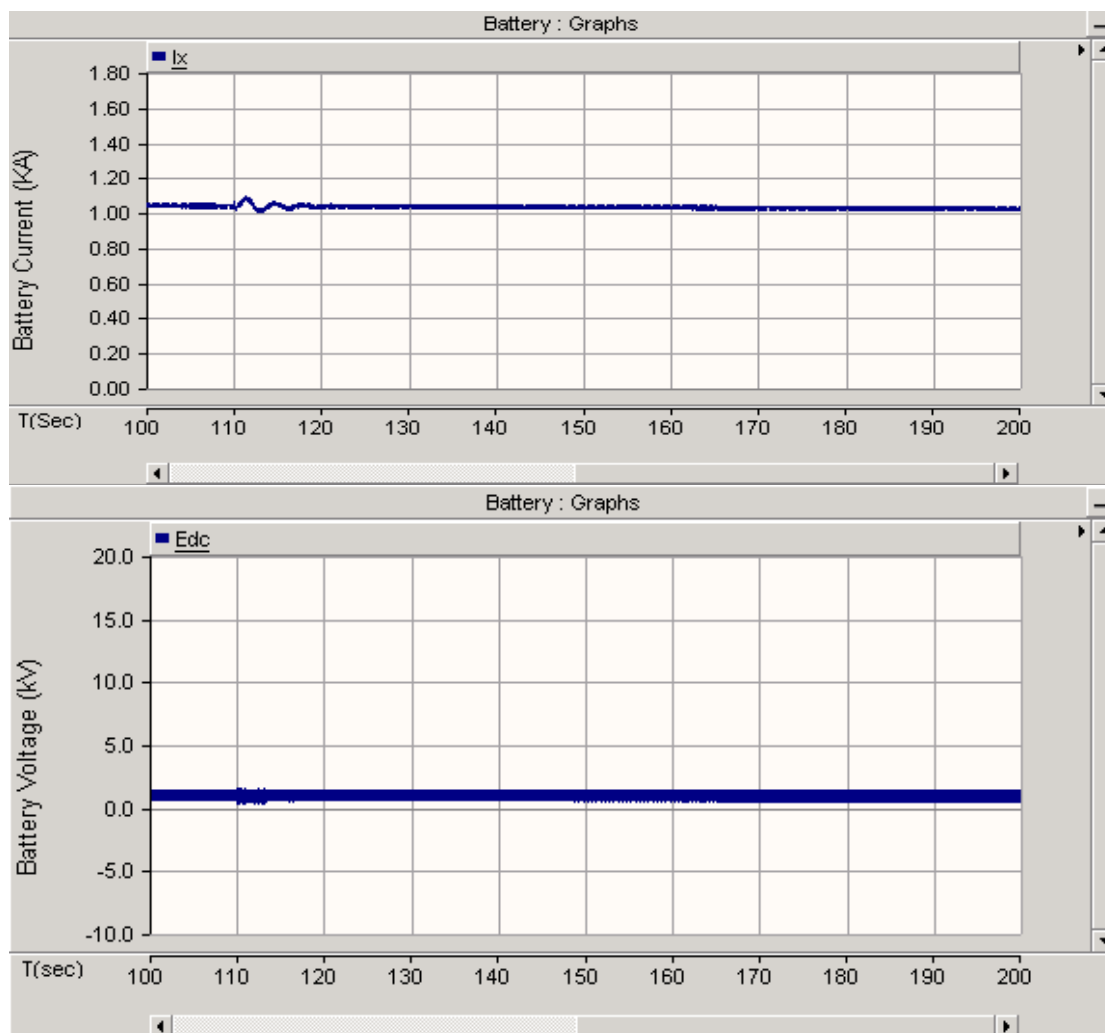


Figure 146. Response at the Battery during Fault (Strong)

A comparison of the preceding graphs show variations and oscillations were controlled better in the strong grid than in the weak grid.

8.5.1.9.3 Grid Connected Wind System

The wind system was connected to the 12.5 kV grid through a variable inductor and a variable resistor. The table modeled in PSCAD was used to vary the R and L values according to the SCR range of 8 to 26. Thus the project settings in PSCAD were adjusted to have multiple runs so that for each run the SCR value changed by reading the corresponding R and L values from the look-up table. The wind system was initially connected to a weak grid for the first run with SCR value of 8 and the corresponding calculated R and L values were 6.5056 ohm and 0.01 H, respectively. The use of the look-up table enabled the change in impedance values and caused the wind system to be connected to a strong grid.

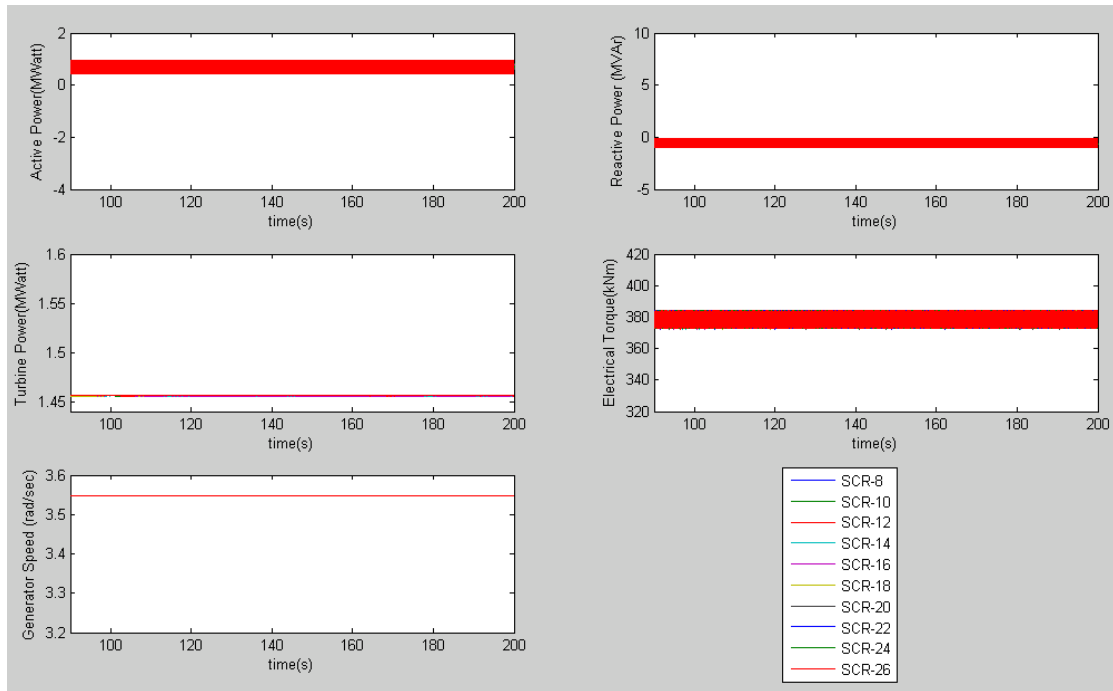


Figure 147. Response of Wind-Connected Grid System

The responses from the wind turbine and grid parameters were observed. The multiple run values were saved to a file and the data was extracted through MATLAB and plotted. The steady state analysis of the wind system connected to grid did not show any variation with the values changed from weak to strong grid. From the above figure, it was observed that active power was 1.7 MW and the turbine power was 1.5 MW. The reactive power was approximately equal to be 0. The mechanical speed was 3.77 rad/sec. The turbine was obtained to be 400 kNm.

The plots in the figure below show the grid response when the wind system was connected to it. The grid rms voltage was found to be 12.2 kV and the grid rms current was 280 A. The active power at the grid was 1.7 MW and the reactive power at the grid was 0.5 Mvar. The response graphs show the legend with the SCR varying from 8 to 26. The legend observes the response of the system transfer from weak to strong grid. Both the grid and wind turbine response did not show much variation in transition from weak to strong grid.

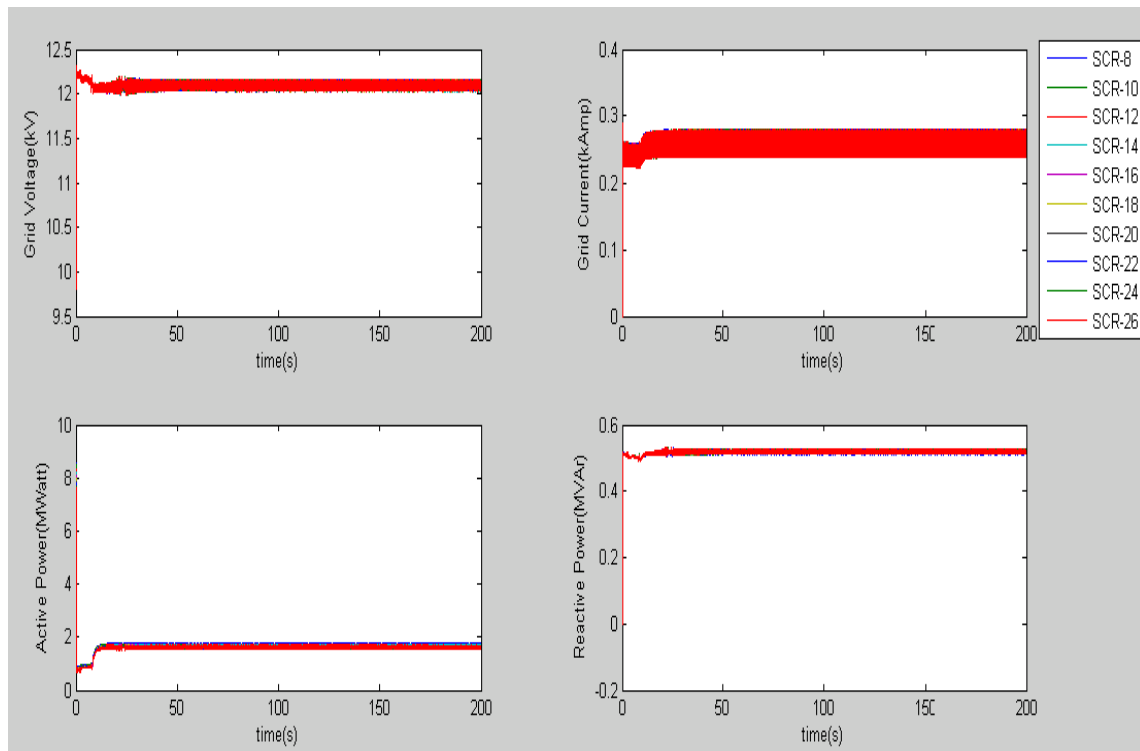


Figure 148. Grid - Voltage and Current (top) & Active and Reactive Power (bottom)

8.5.1.9.4 Grid Connected Wind System-Fault

A three-phase fault to ground was applied near the grid to the system as shown below. The transient state analysis of the wind connected to grid showed variations when the values were changed from weak to strong grid through the SCR.

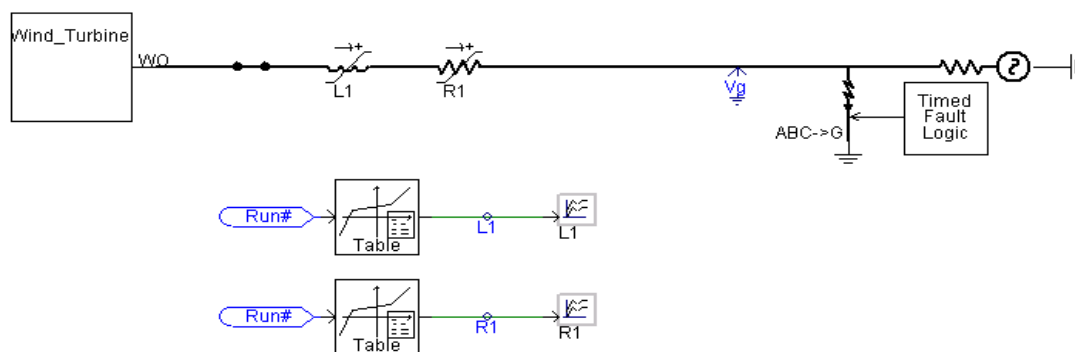


Figure 149. Wind-Connected Grid System with Fault Applied near the Grid

The active power, reactive power and electric torque parameters in the figure below were found to have unsteady response when simulated for a weak grid condition at low SCR of 8. While the response was

stabilized as the SCR increased and simulated for a strong grid with an SCR of 26. The three-phase fault was applied between 110 and 113 seconds. The plots below show the grid response when the wind system was connected to it during transient conditions. The grid rms voltage was found to be 12.2 kV and decreased during the fault. The grid rms current was 280A initially but increased during the fault. The active power at the grid reduced and the reactive power at the grid increased during the fault and settled after oscillations.

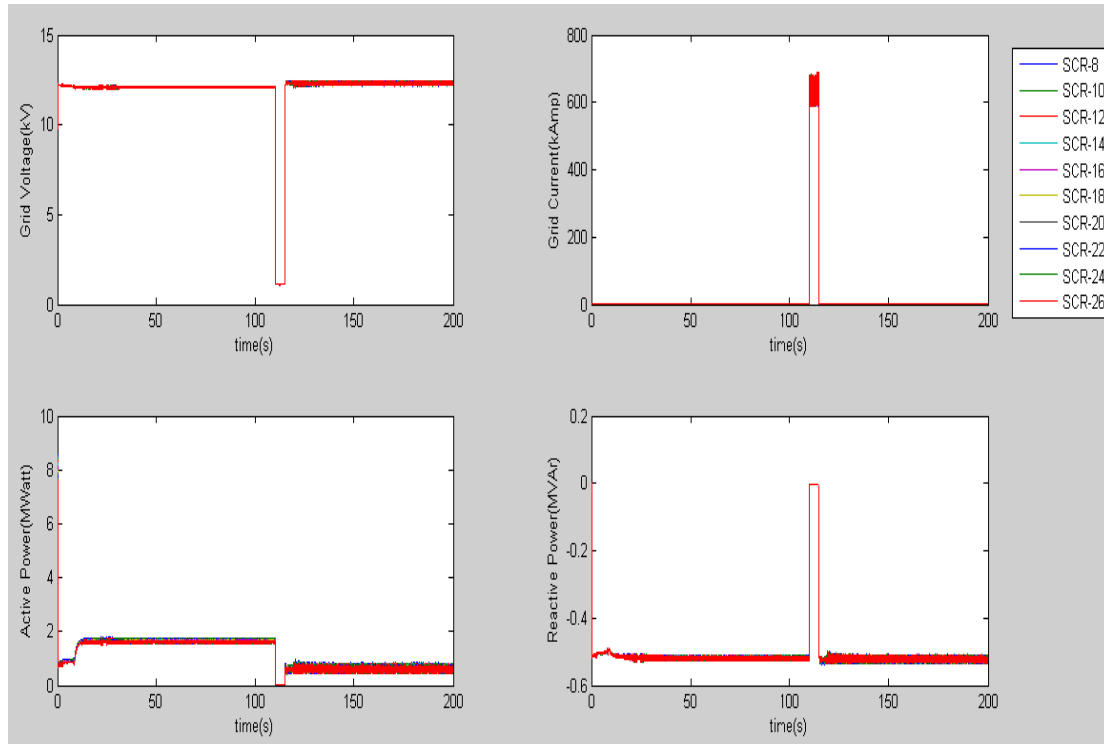


Figure 150. Grid - Voltage and Current (top) & Active and Reactive Power (bottom)

8.5.1.9.5 The Complete System

The battery system and the wind system were connected to the 12.5 kV grid through a variable inductor and a variable resistor as shown below. Look-up tables and project settings were made the same as described in previous sections and were adjusted to have multiple runs so the SCR values changed and took the corresponding R and L values through the multiport switch. The battery system and wind system were initially connected to a weak grid for the first run with SCR value of 8 and the corresponding calculated R and L values were 6.5056 ohm and 0.01 H, respectively.

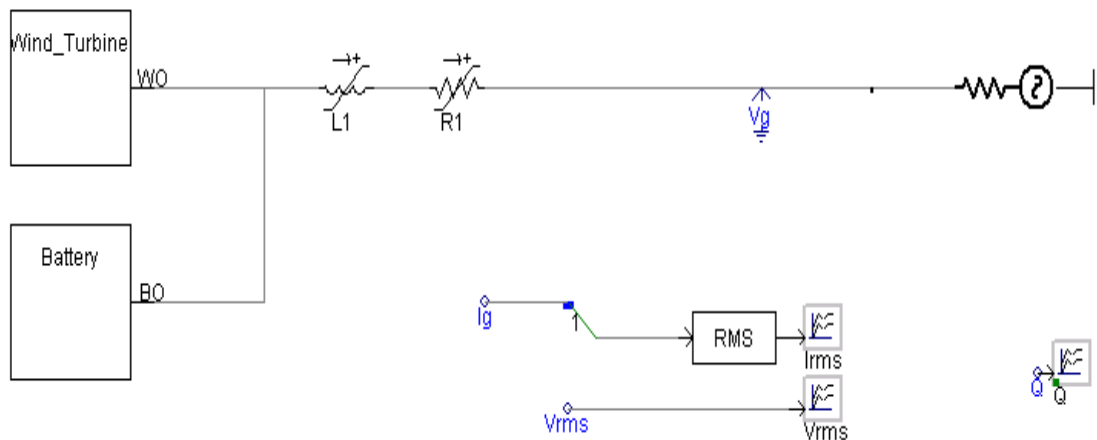


Figure 151. Complete System with Battery, Wind Connected to Grid

The response of the battery, wind turbine and grid parameters were observed. The multiple run values were saved to a file and the data was extracted through MATLAB and plotted as shown below. The steady state analysis of the battery and wind connected to grid did not show any variation with the values changed from weak to strong grid. The battery discharge was observed in the battery voltage to be 960 V until time for 200 seconds and continued to discharge until the nominal voltage at 3600 seconds. The battery current was found to be settling at 1100 A.

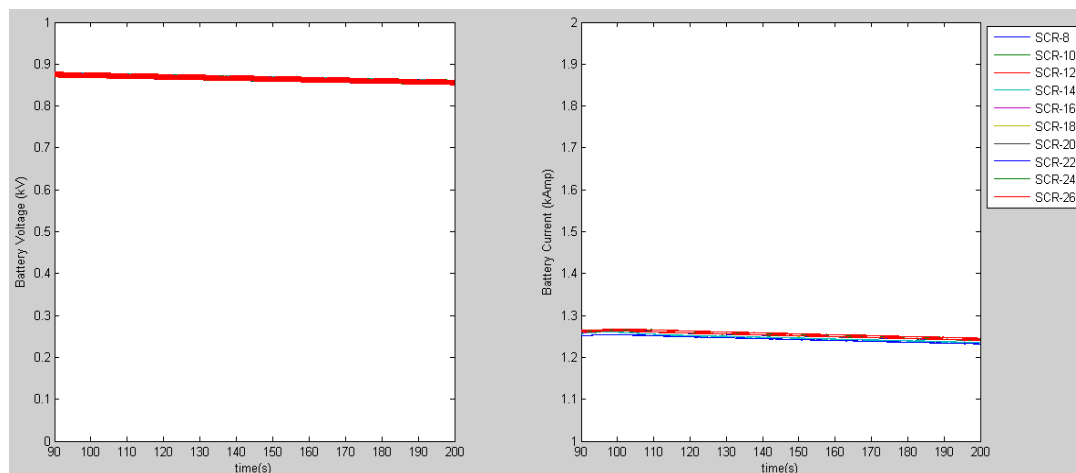


Figure 152. Battery Response when Wind Connected to Grid

From the following figure it was observed that the active power was 1.3 MW and the turbine power was ≈ 1.5 MW. The reactive power was approximately equal to 0. The mechanical speed was 3.77 rad/sec. The turbine was obtained to be 400kNm. The oscillations at the initial conditions were due to the addition of filters to the wind system.

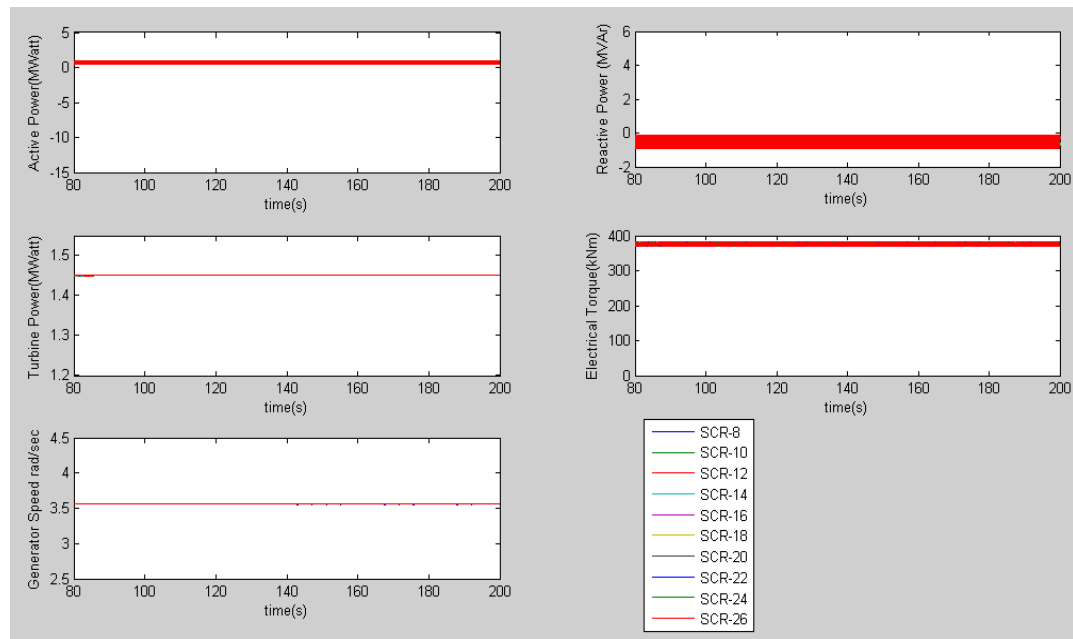


Figure 153. Wind Response when Battery Connected to Grid

The plots below show the grid response when battery and wind system were connected to it. The grid rms voltage was found to be 12.0 kV and the grid rms current was 350 A. The active power at the grid was 2.7 MW and the reactive power was 0.2 Mvar. The grid, wind and battery response did not show much variation in transition from weak to strong grid.

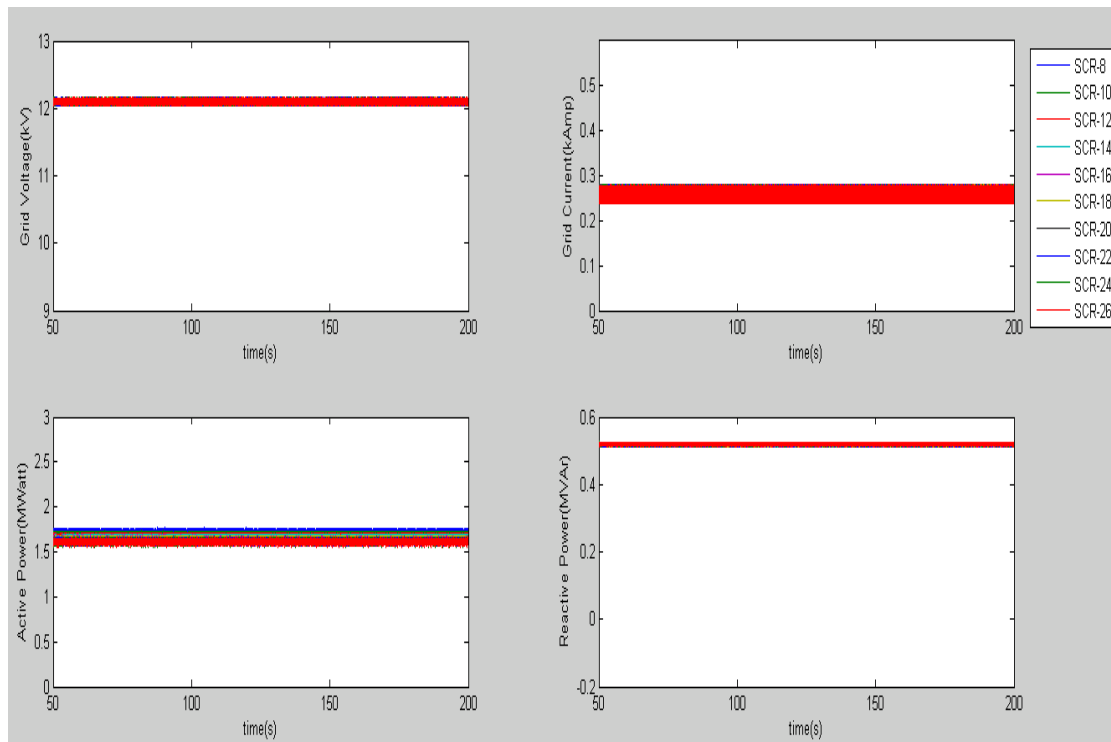


Figure 154. Grid Response when Battery and Wind Connected to It

8.5.1.9.6 The Complete System-Fault

A three-phase fault to ground was applied near the grid to the system as shown below. The transient state analysis of the battery, wind connected to grid showed variations with battery, wind and grid components when the impedance values were changed from weak to strong grid through the SCR.

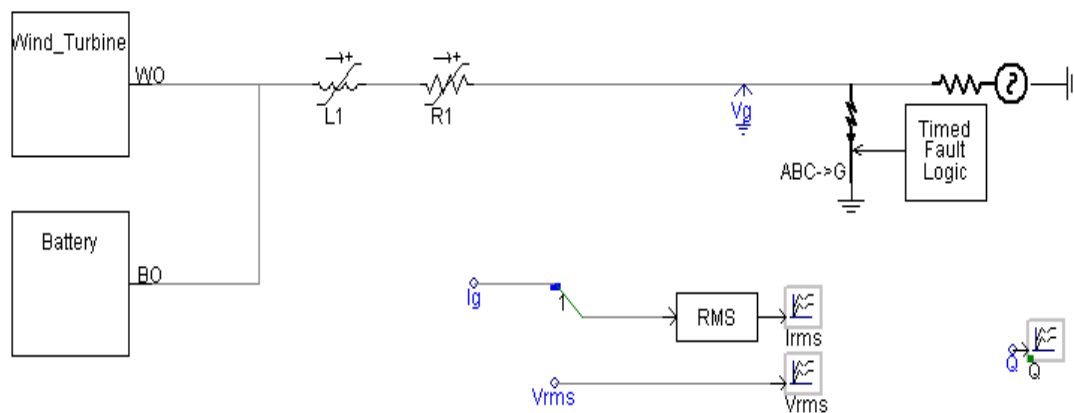


Figure 155. Complete System with Three-Phase Fault

The battery discharge characteristic for weak grid condition in the figure below was observed to be 1000 V (“Edc”) until time at 110 seconds and had a reduction in voltage level during the three-phase fault period applied between 110 -113 seconds. The voltage resumed back after the fault period and attained its steady state condition. The battery current was found to be increasing to 2500 A during the fault period and settled after the fault at 1100 A.

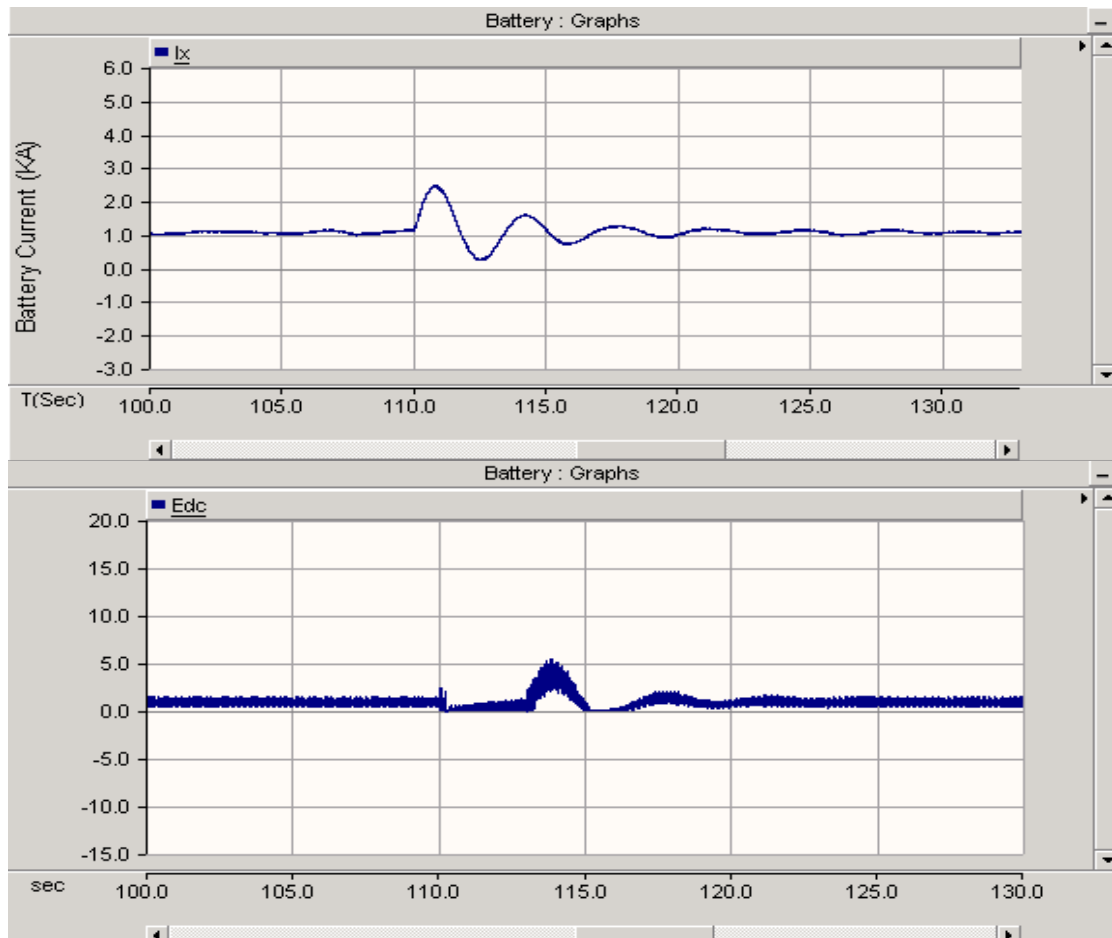


Figure 156. Battery Response when Wind Connected to Grid with Fault (Weak)

The battery discharge characteristic for weak grid condition shown below was observed to be 1000 V (“Edc”) until time at 110 seconds and had a reduction in voltage level during the three-phase fault period applied between 110-113 seconds. The voltage resumed back after the fault period and attained its steady state condition. The battery current was found increasing to 1800 A during the fault period and settled after the fault at 1100 A.

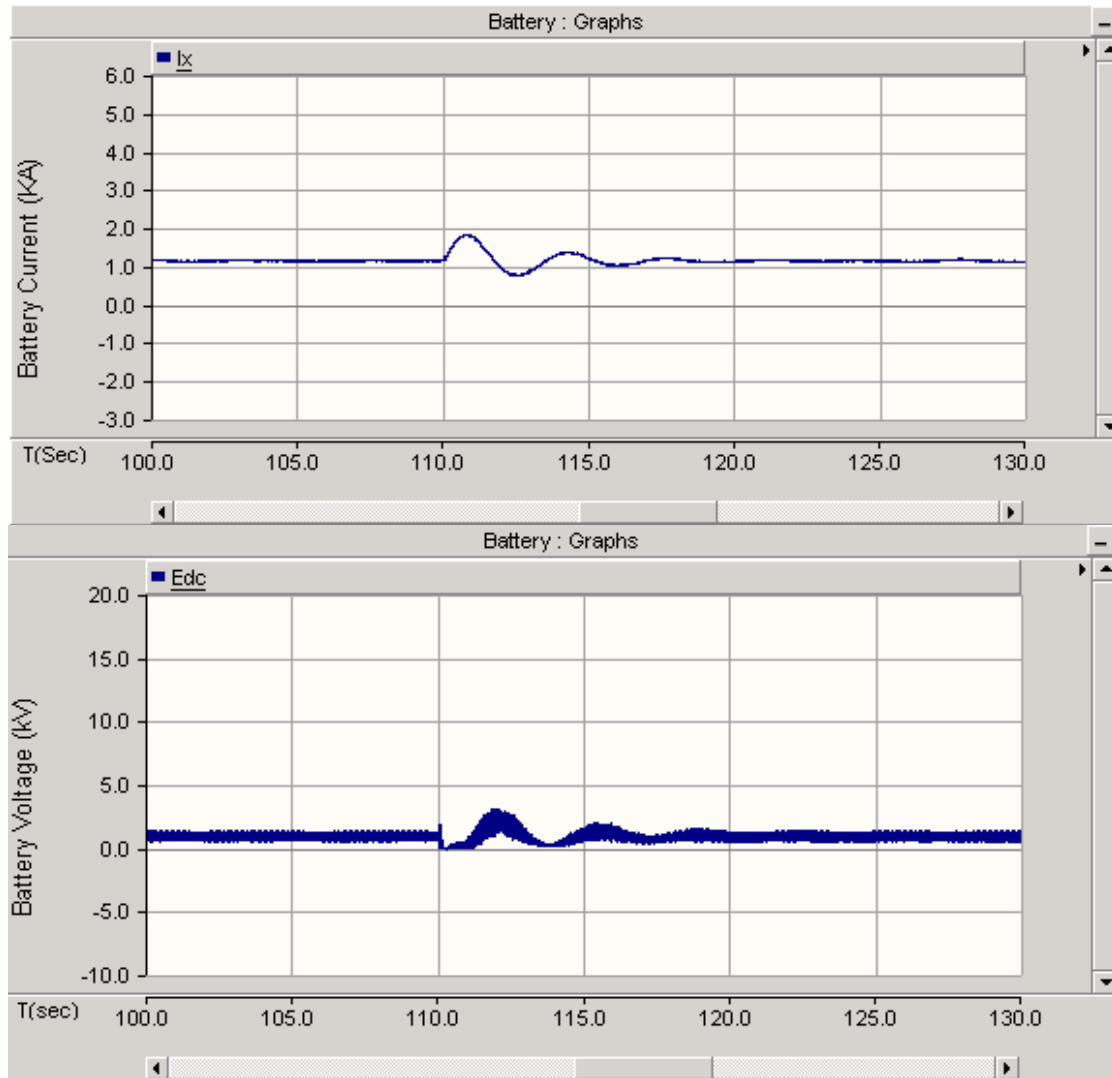


Figure 157. Battery Response when Wind Connected to Grid with Fault (Strong)

From the above results, it was observed that the weak grid had high overshoot for 1 second in current and voltage which was not so with the strong grid where the overshoot during the fault was comparatively less.

The transient state analysis of the wind turbine connected to grid along with battery showed variations when the values were changed from weak to strong grid through the SCR. The active power, reactive power and electric torque parameters in the figure to the left were found to have unsteady response when simulated for a weak grid condition at low SCR of 8. While the response was stabilized as the SCR increased and was simulated for a strong grid with SCR of 26 in the figure to the right. The three-phase fault was applied between 110 and 113 seconds.

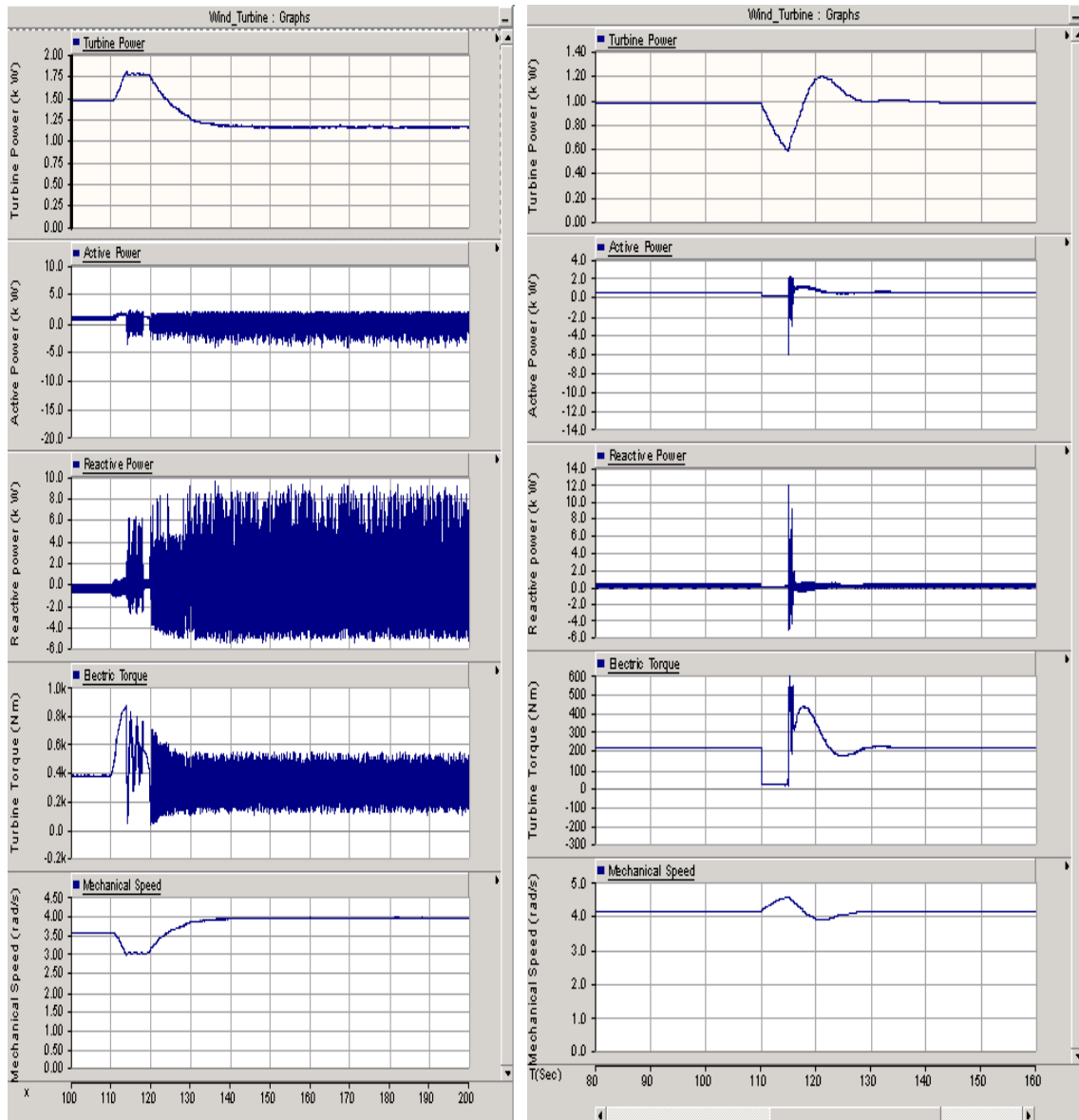


Figure 158. Wind Response - Battery to Grid with Fault – Weak (left), Strong (right)

The plots show the grid response when battery and wind system were connected to it during transient conditions. The grid rms voltage was found to be 12.2 kV and decreased during the fault. The grid rms current was 350 A initially but increased during the fault. The active power at the grid reduced and the reactive power at the grid increased during the fault and settled after oscillations.

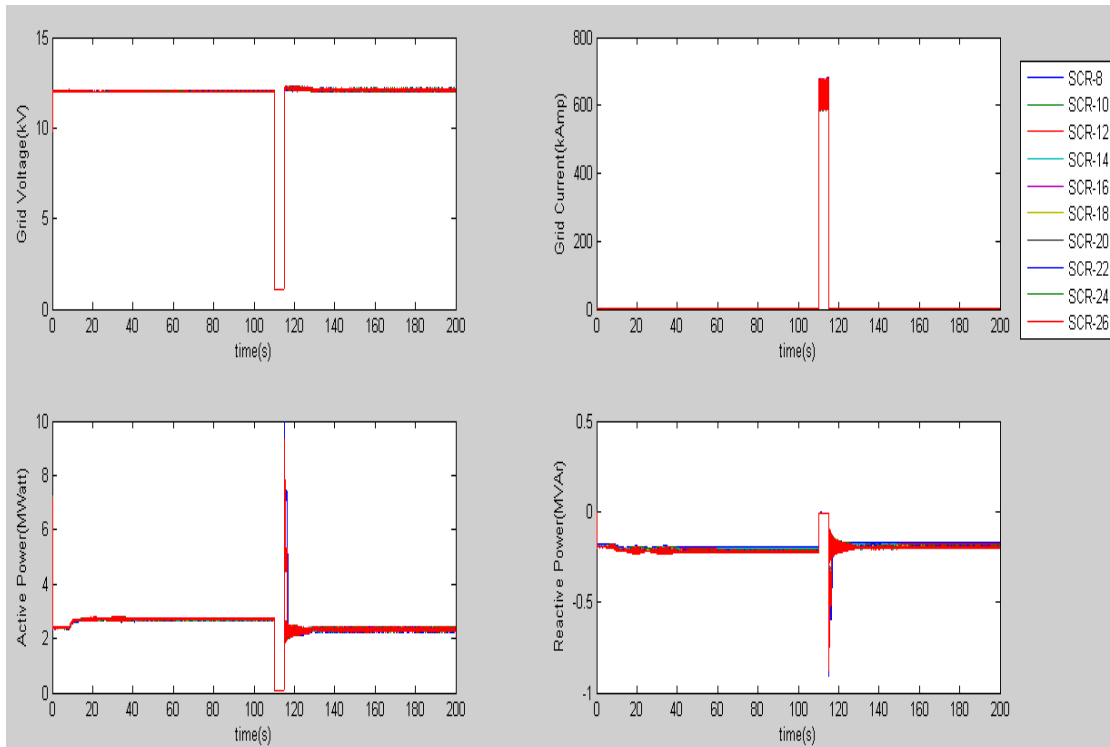


Figure 159. Grid Response when Battery and Wind Connected to It

In the above simulations, the performance of the equivalent 1MWh battery circuit and wind turbine parameters was analyzed with the weak/strong concept. It was observed from the above analysis that the controls play a vital role and require optimization of the battery and wind parameters so that when the system reaches unstable conditions such as transient fault conditions in a weak grid, the effective controls and addition of control filters would resume the stable operation of the grid. In this project, the addition of the 1 MW BESS to the wind connected grid system cannot help the system as the controls are separate for both battery and wind. However, once the fault conditions, voltage dip or frequency fluctuations as well as low power generation at the wind turbines are sensed by the controls in the battery system, the battery can respond to actively support the system.

8.5.2 Analysis of Wind Turbine Data

Beginning in April 2013, the Team began collecting and analyzing wind turbine data from the resources at the Reese Technology Center. This included the Alstom turbine, and later the Gamesa turbines. The figures below show the daily average data of the power sent to the local substation in kW, the turbine production, during the months when the battery was tested in 2014 from June to October. The power production and cost analysis is done to provide a baseline for the performance of economic dispatch analysis to determine optimal battery charge/discharge schedules.

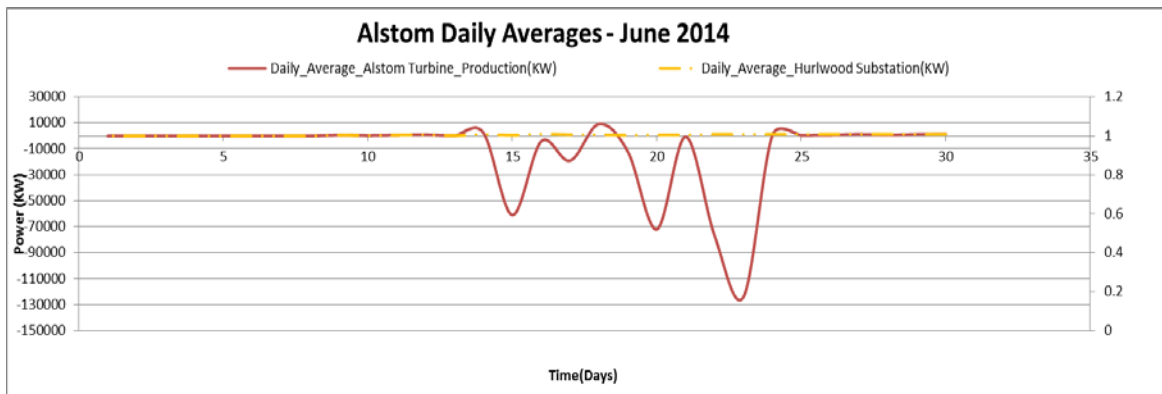


Figure 160. Alstom-Daily Averages for June 2014

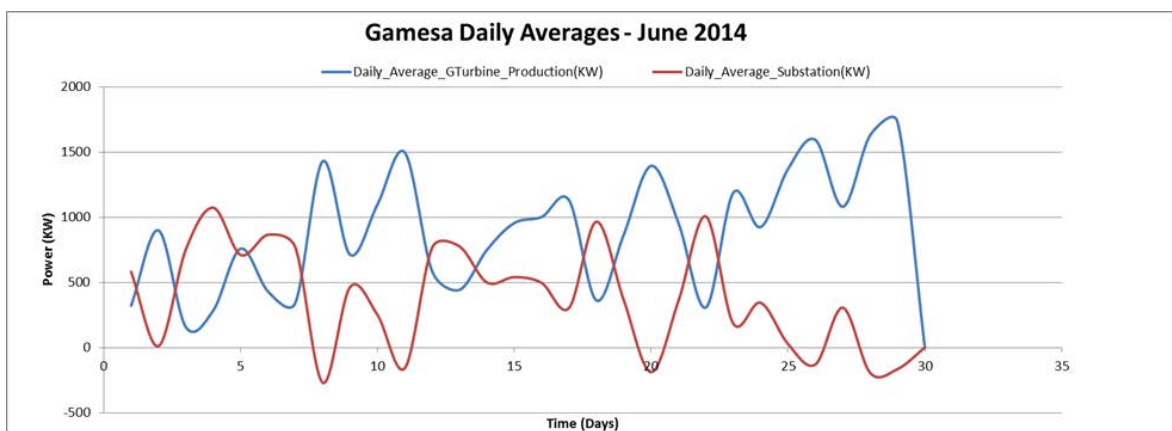


Figure 161. Gamesa-Daily Averages for June 2014

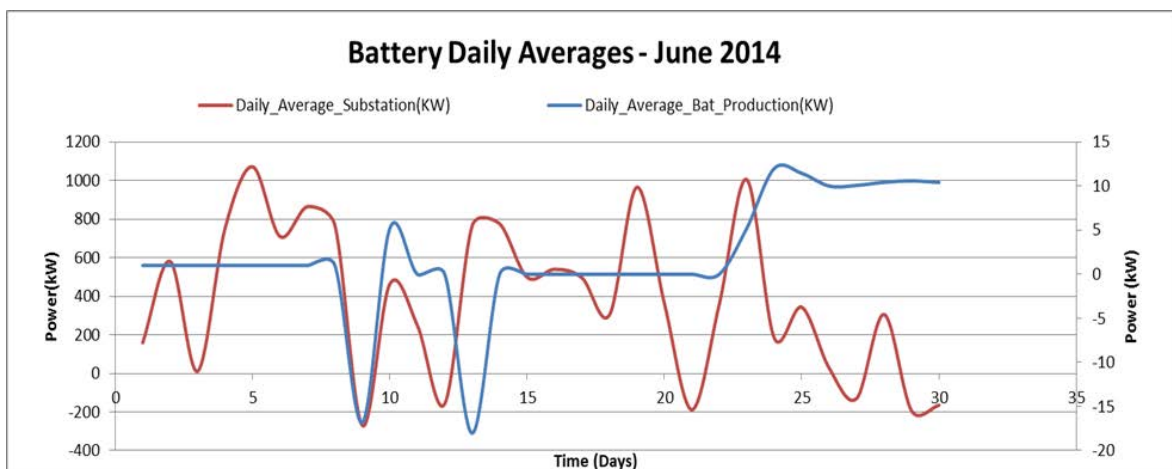


Figure 162. Battery-Daily Averages for June 2014

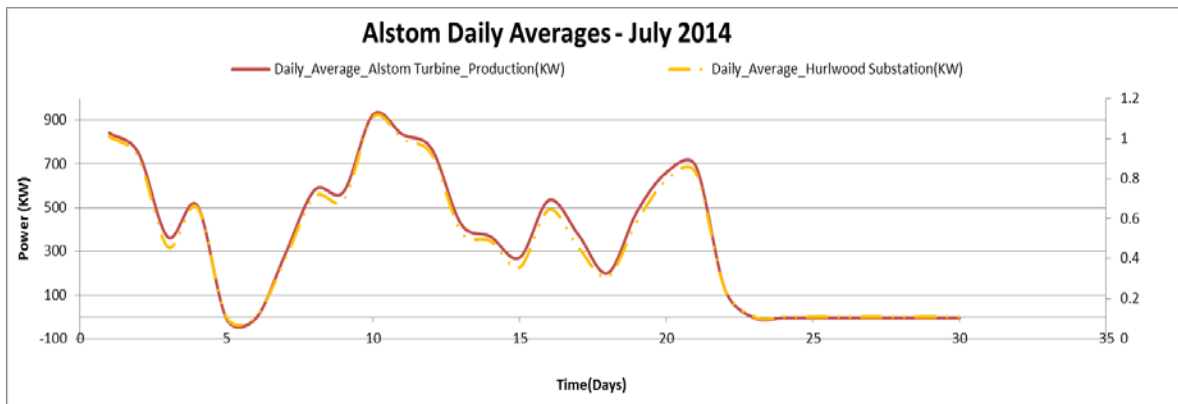


Figure 163. Alstom-Daily Averages for July 2014

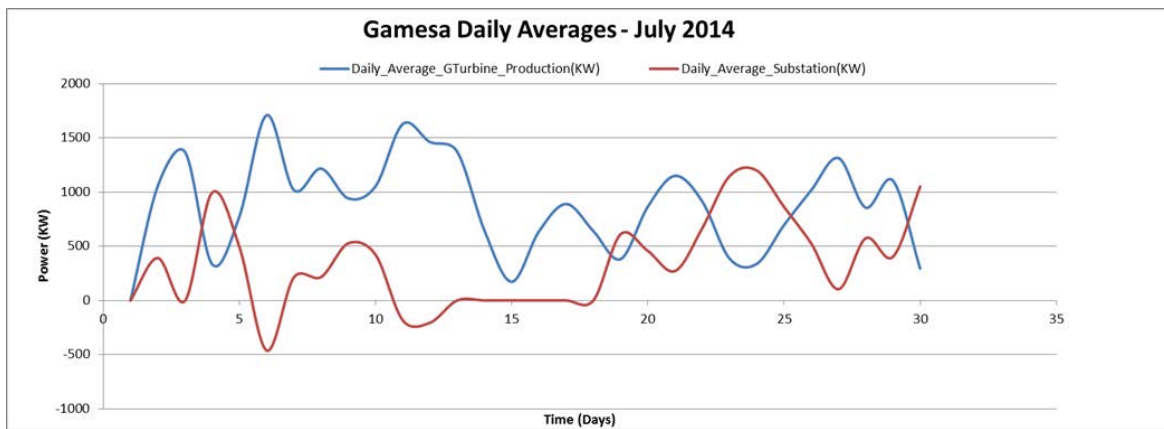


Figure 164. Gamesa-Daily Averages for July 2014

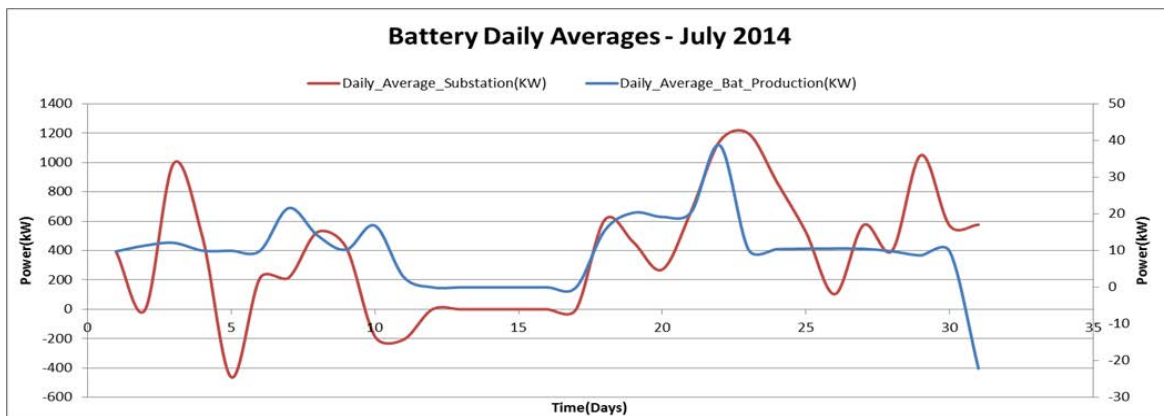


Figure 165. Battery-Daily Averages for July 2014

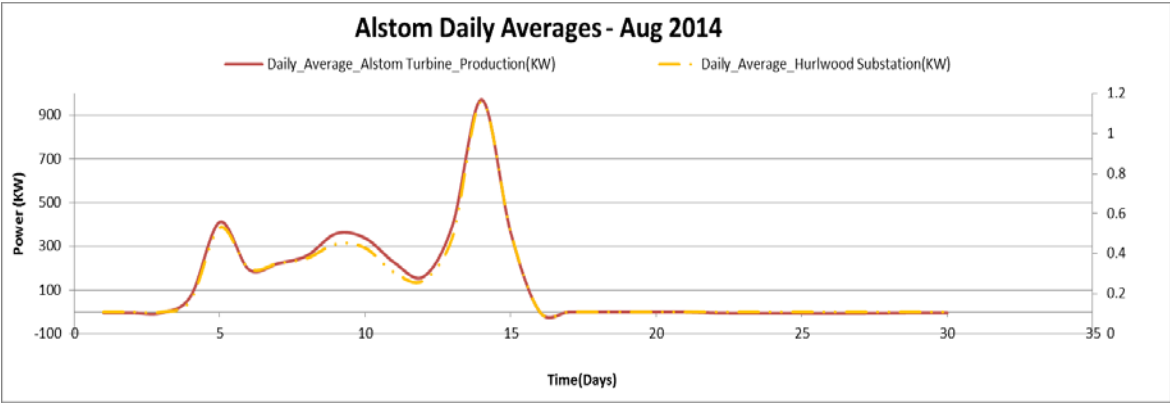


Figure 166. Alstom-Daily Averages for August 2014

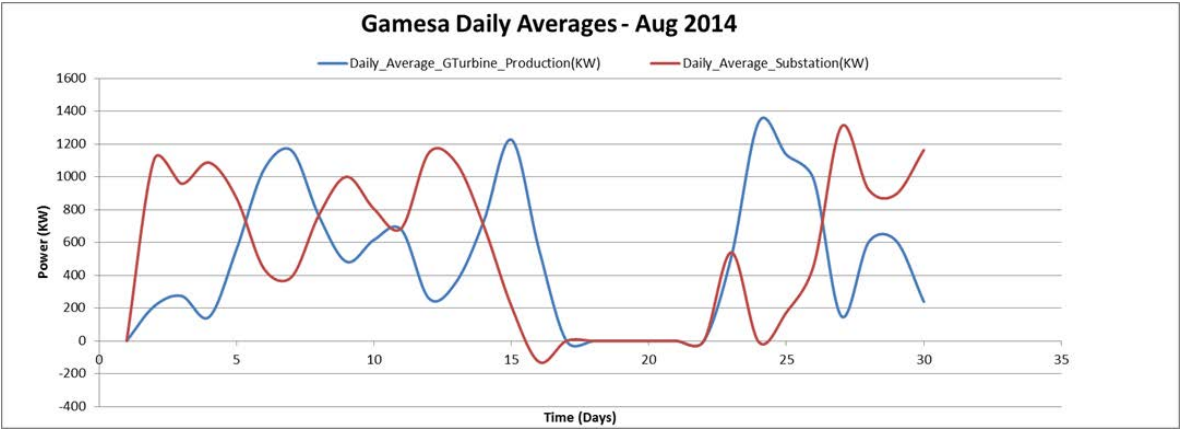


Figure 167. Gamesa-Daily Averages for August 2014

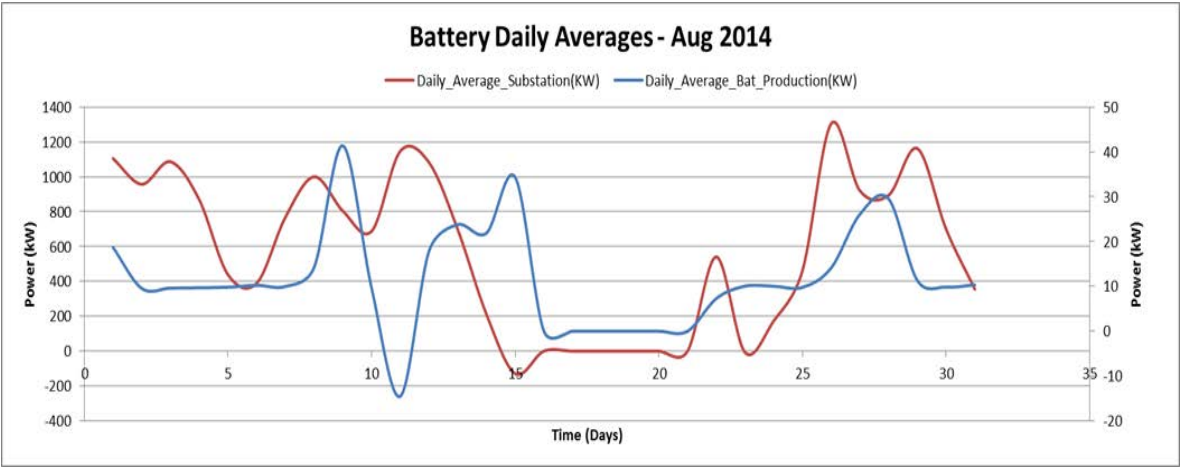


Figure 168. Battery-Daily Averages for August 2014

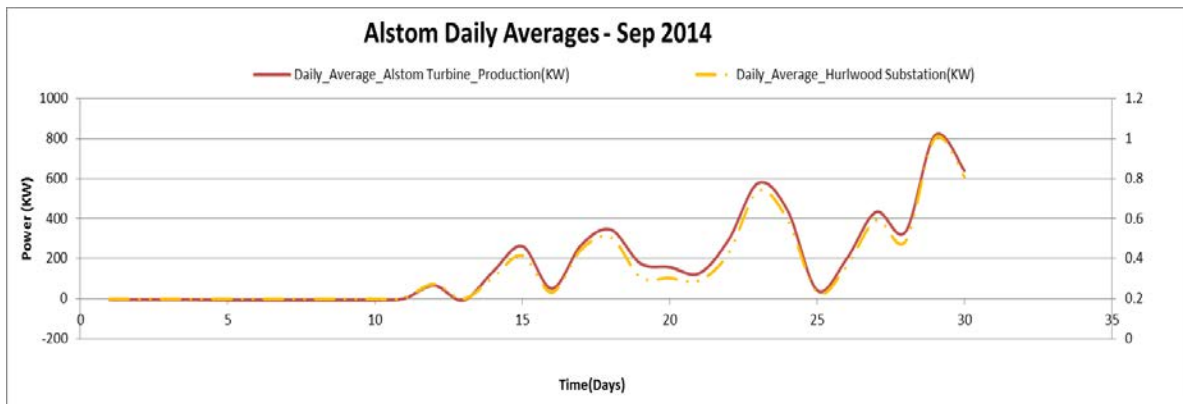


Figure 169. Alstom-Daily Averages for September 2014

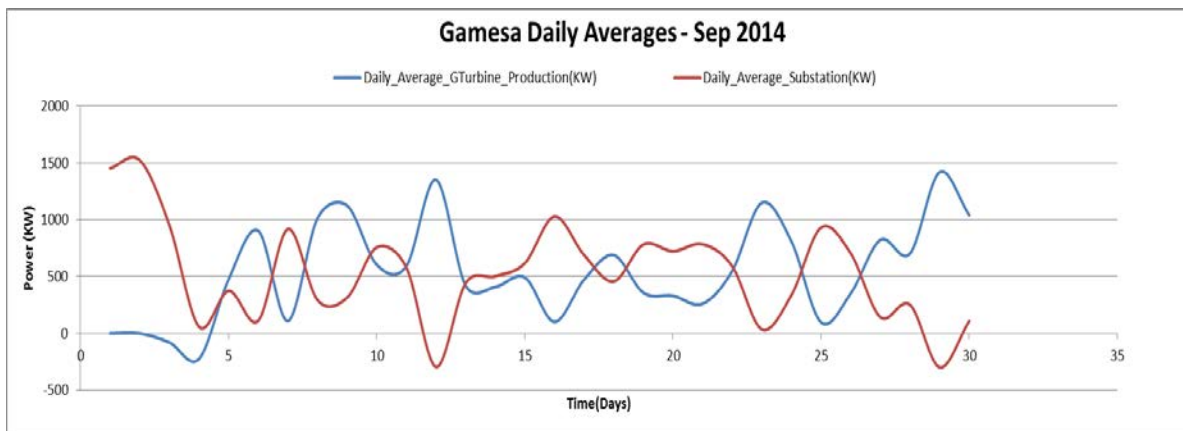


Figure 170. Gamesa-Daily Averages for September 2014

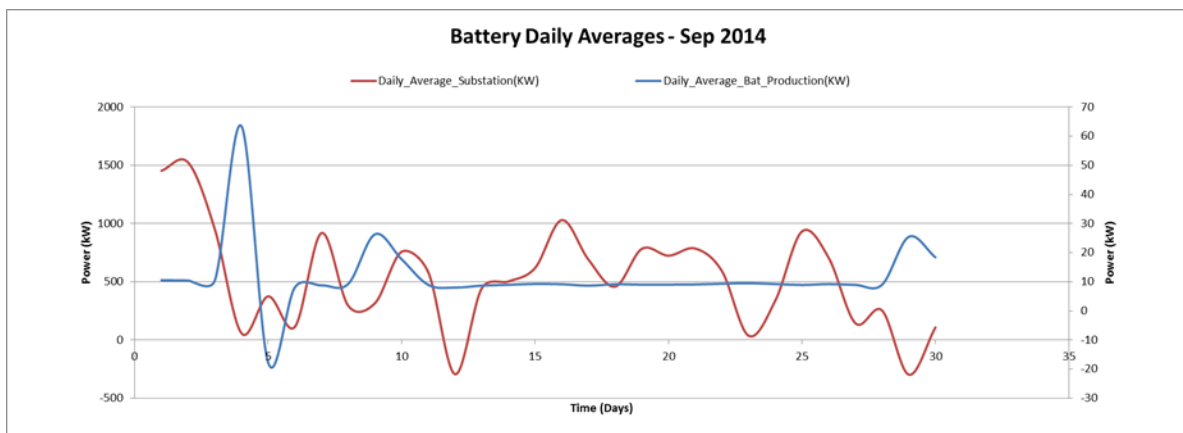


Figure 171. Battery-Daily Averages for September 2014

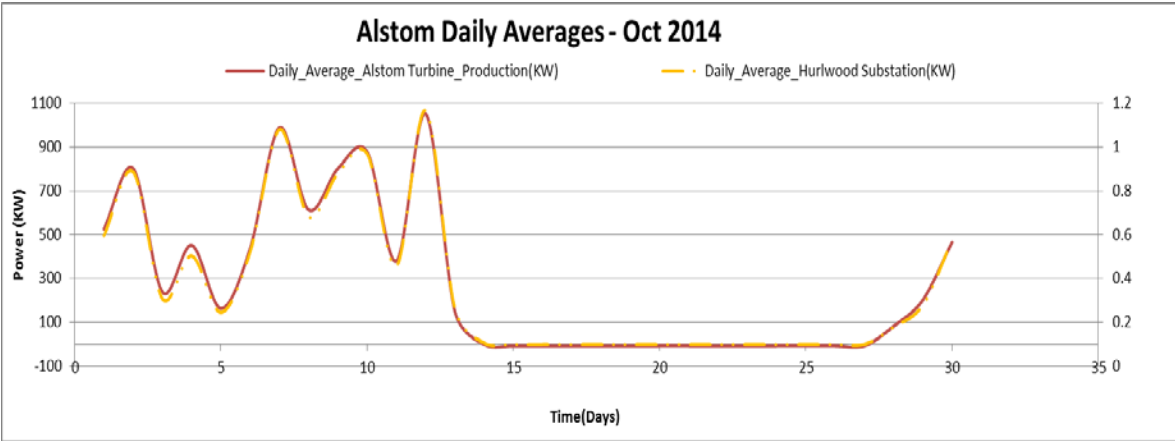


Figure 172. Alstom-Daily Averages for October 2014

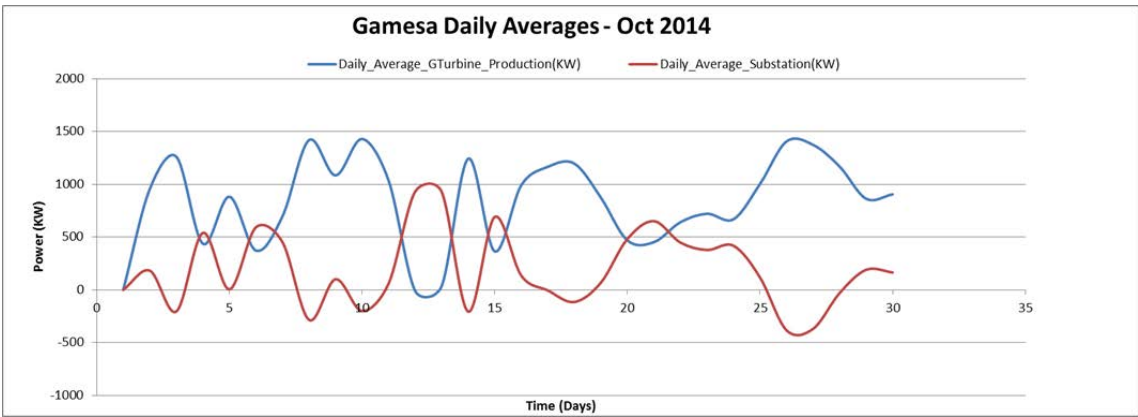


Figure 173. Gamesa-Daily Averages for October 2014

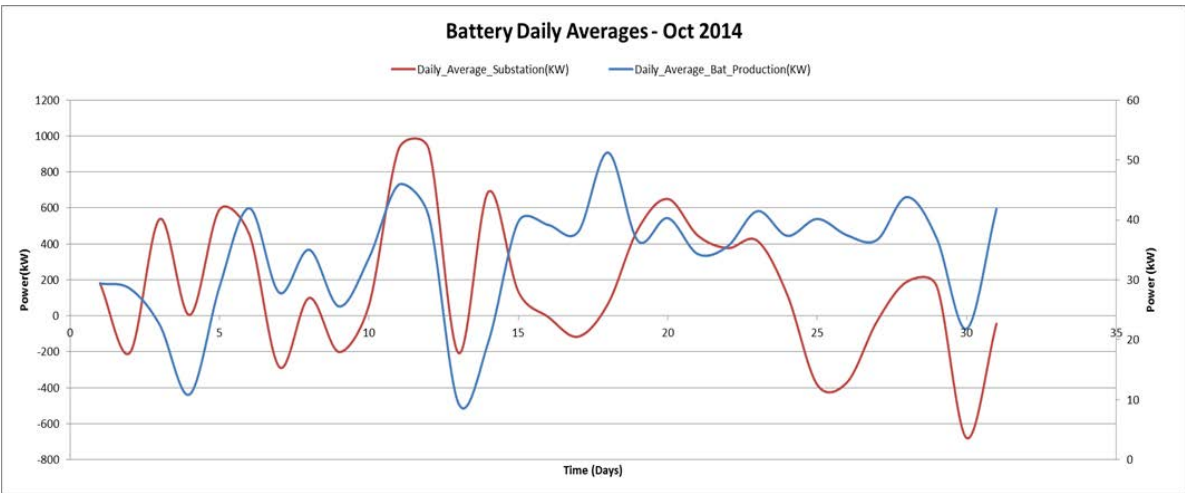


Figure 174. Battery-Daily Averages for October 2014

8.5.3 Battery Testing in 2014

The BESS testing was performed during 2014. The basic approach for testing, describing such aspects as time of year, time of day, and duration, is outlined in the table below. This is then followed with test results.

Table 31. Overall BESS Testing Approach

Number	Test Type	Time of Year	Time of Day	Duration
1	Ramping	All year	Night	Seconds
2	Demand Response	Summer (May to August)	Day	Minutes to hours depending on peak loads
3	Frequency Support	All year	Night	Seconds
4	Wind Speed Drop	All year	When drop occurs	Seconds to minutes
5	FRRS	All year	When needed	Until frequency recovers

8.5.3.1 Ramp Support

The battery was ramped to an SOC of 50%. It was charged at a rate of 250 kW for 15 minutes, or until reaching 50% SOC. Then charge rate was increased to 500 kW for 15 minutes, or until reaching 50% SOC. Later was the charged to 750 kW and 1000 kW for 15 minutes, or until 50% SOC.

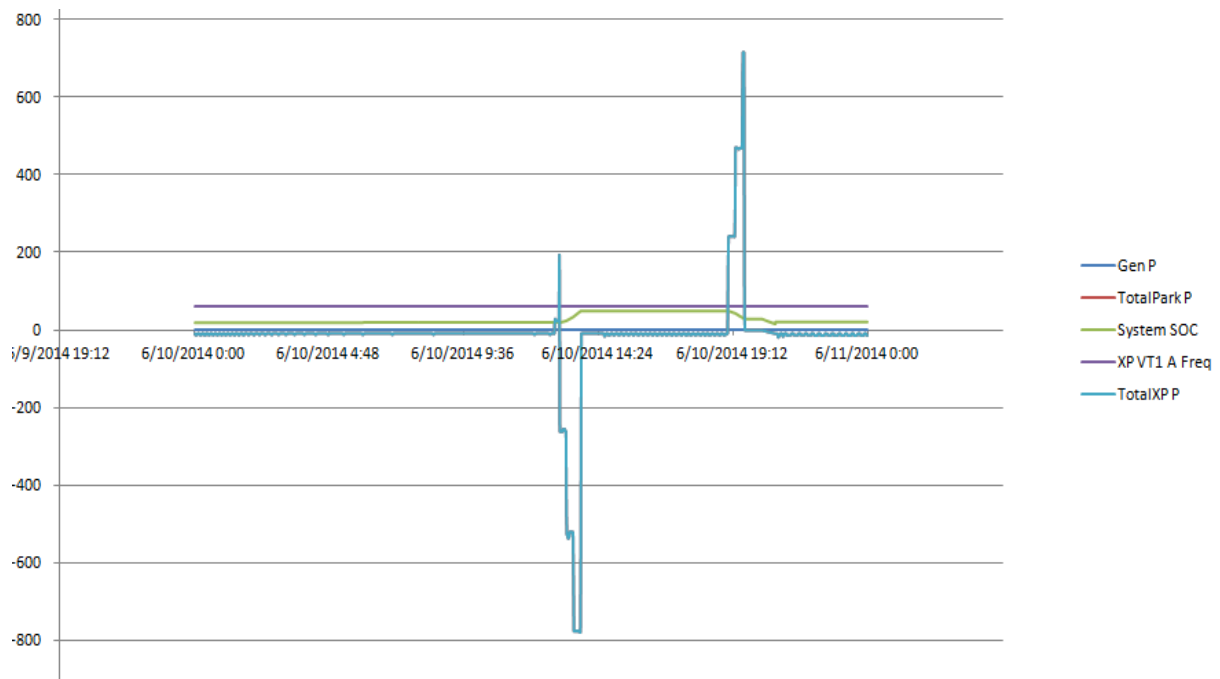


Figure 175. Ramp Support Test on June 9, 2014

8.5.3.2 Frequency Response

When the frequency went below 60 Hz, the battery ramped up to supply the required power and meet the demand. The black line is frequency (dead band is set to 59.70Hz), the orange line is system SOC, and the blue line is kW output (with a maximum setting at 400 kW). The following figure demonstrates that the system responded very well.

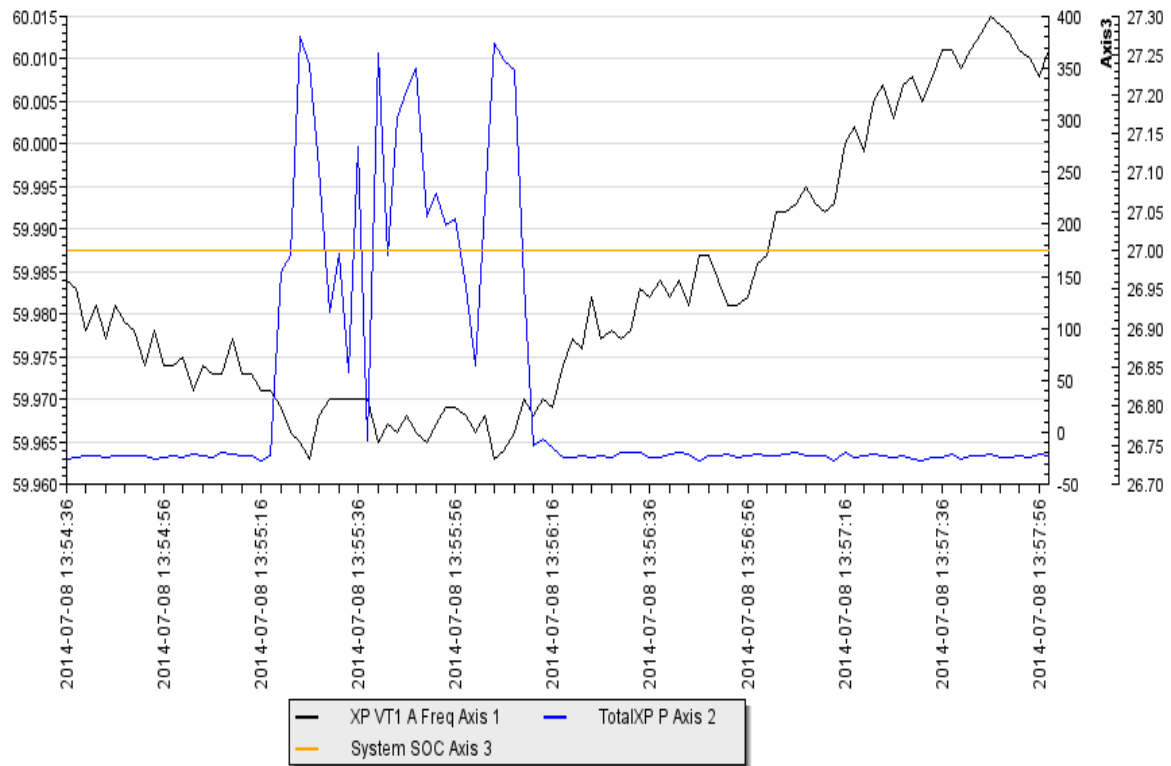


Figure 176. Under Frequency Event on July 08, 2014

8.5.3.3 Wind Ramping/Load Leveling

The figures below show the battery following the wind profile according to the power generation and load demand. The green trace is the wind output for TTU-04 (Gamesa Wind Turbine) and dark blue trace is the output of the battery.



Figure 177. Load Leveling During August 2014

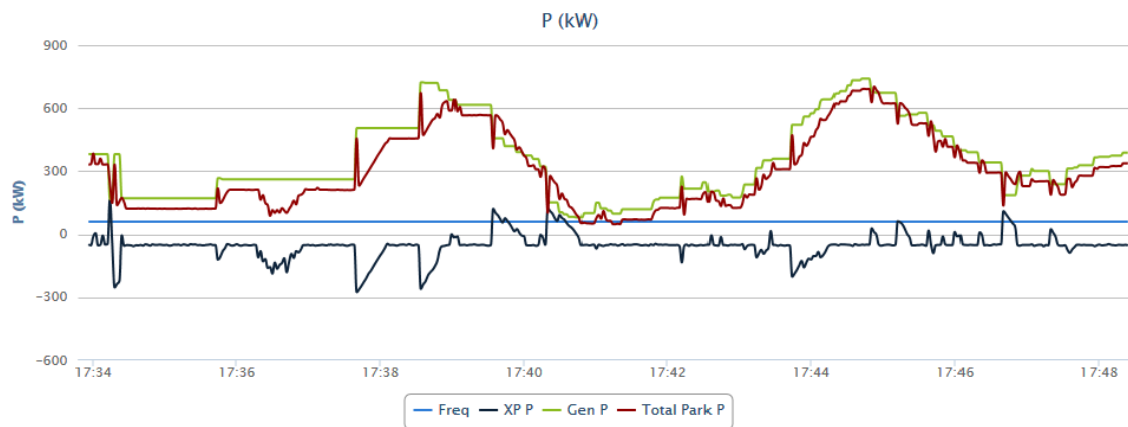


Figure 178. Load Leveling During October 2014

8.5.3.4 Demand Reduction

The following figure shows the battery providing peak reduction/demand shaving. Younicos charged the battery to 90% SOC and discharged the battery at 1000 KW for 53 minutes or when the SOC reached 10%.

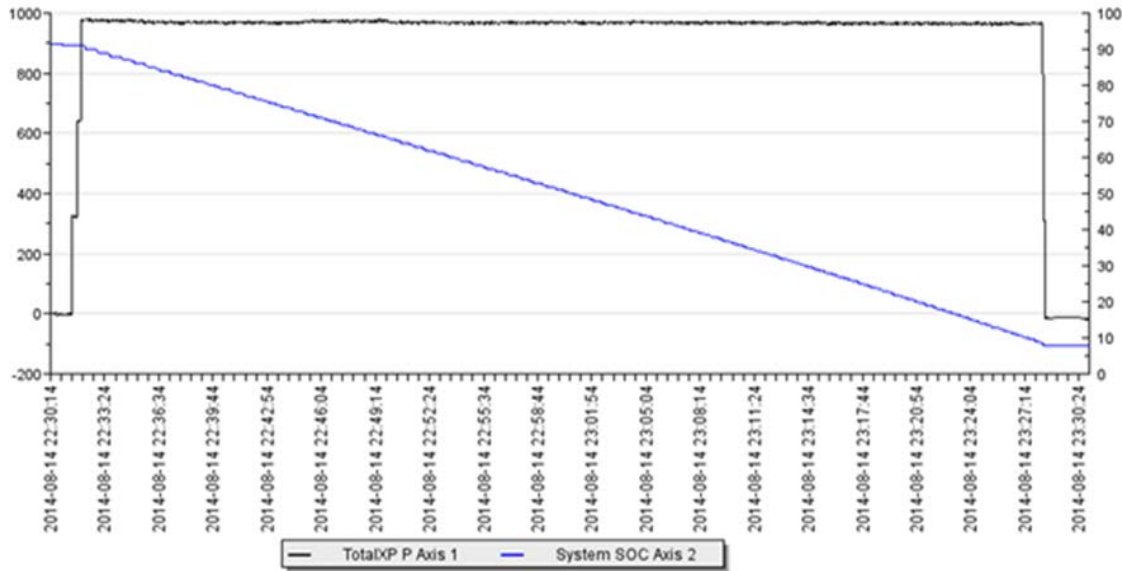


Figure 179. Demand Reduction during August 2014

8.5.4 Economic Analysis

There are a wide range of uses for a BESS, from local voltage stability to actually providing power to homes and businesses in rural areas. However, the solutions are expensive and the benefits depend greatly on the functionality that can be delivered. Hence an economic analysis is essential to decide the feasibility of any given storage system, and there are various factors that need to be considered when performing the analysis. One of the most important factors is to determine the optimal time to charge and discharge the battery. The battery needs to be charged when demand and electricity cost are low and discharged when demand and electricity cost is high. It all depends on the purchase price of energy required to charge the battery and alternatively the sale price of energy to sell the excess power. The next important factor is to consider the load/demand and supply of the battery power or the available renewable power in an economic way considering the past, present and future cost analysis. The following algorithm provides an analysis of a solution for determining appropriate battery actions to increase profit while considering depreciation of the battery life. The algorithm for the analysis is laid out as follows with reference to the flowchart in the figure below that. The objectives are defined and the evaluations are done taking the available data, risks and assumptions to decide the best optimal solution for the economic performance of the battery. The following steps are considered in the algorithm.

1. Maximize operation and minimize cost
2. Optimal time to charge (energy purchase shifting), load profile (daily, annual)
3. Economic evaluations:
 - Cost analysis
 - Cost-minimization analysis
 - Cost-effective analysis

- Cost-benefit analysis
- Cost-utility analysis
- 4. Data, assumptions, and risks
 - BESS round trip efficiency
 - BESS life span in cycles
 - Rated power
 - Rated discharge time and capacity
 - Renewable energy availability
- 5. Cost
 - BESS capital cost
 - BESS operating cost (energy-related/not energy-related)
 - Differential cost of energy
 - LMP
 - Present and future values
- 6. Compare, decide and conclude

The algorithm is based on the various input data such as: battery production (MW), wind production (MW), load (kW), local marginal price (LMP), and frequency response of the system, as shown in the figure below. The inputs are fed into the algorithm as shown and then the battery performance or its recommended action is obtained as outputs.

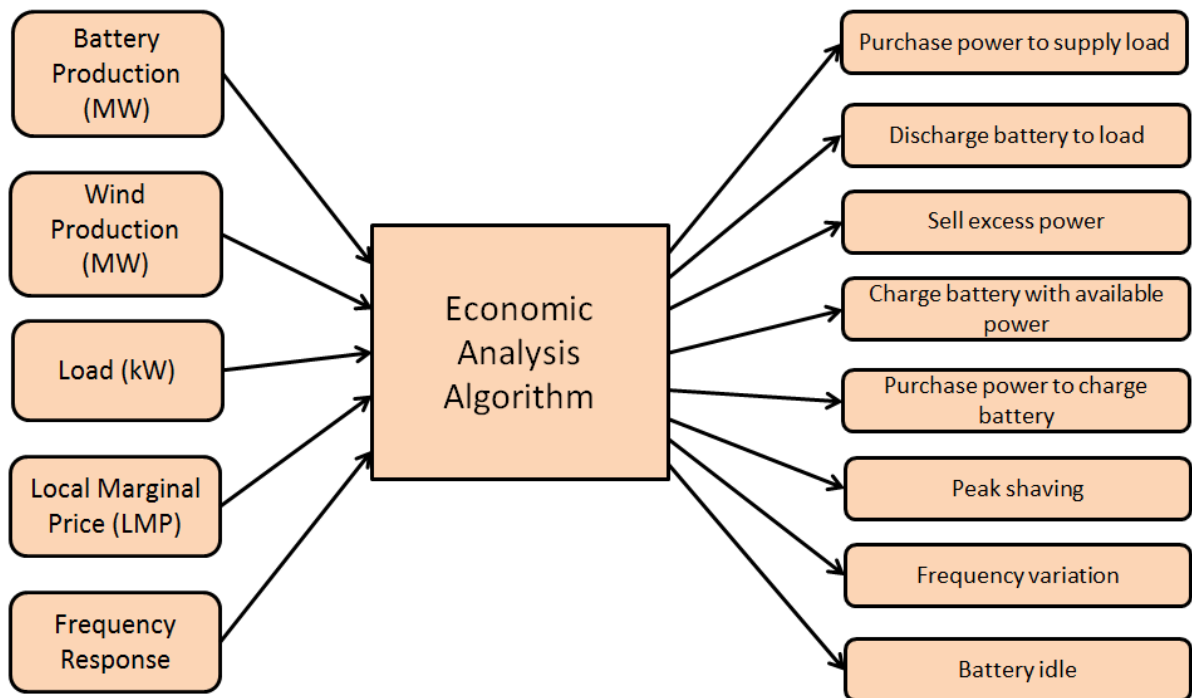


Figure 180. Economic Analysis Block Diagram

The flowchart was laid out for the analyses of the battery operation for its optimal performance and economic benefits.

8.5.4.1 Algorithm for Power Supply Less Than Load

The generalized flowchart in the figure below considers the condition when the wind power is greater than the load and when it is less than the load. For both conditions, the battery state of charge is considered as well as the past, present and future cost data. The algorithm includes decisions at various conditions and then terminates with the stop condition at the end. The algorithm is run by a Python program that reads values from an Excel file with updated data every 15 minutes, and then executes its logic programming.

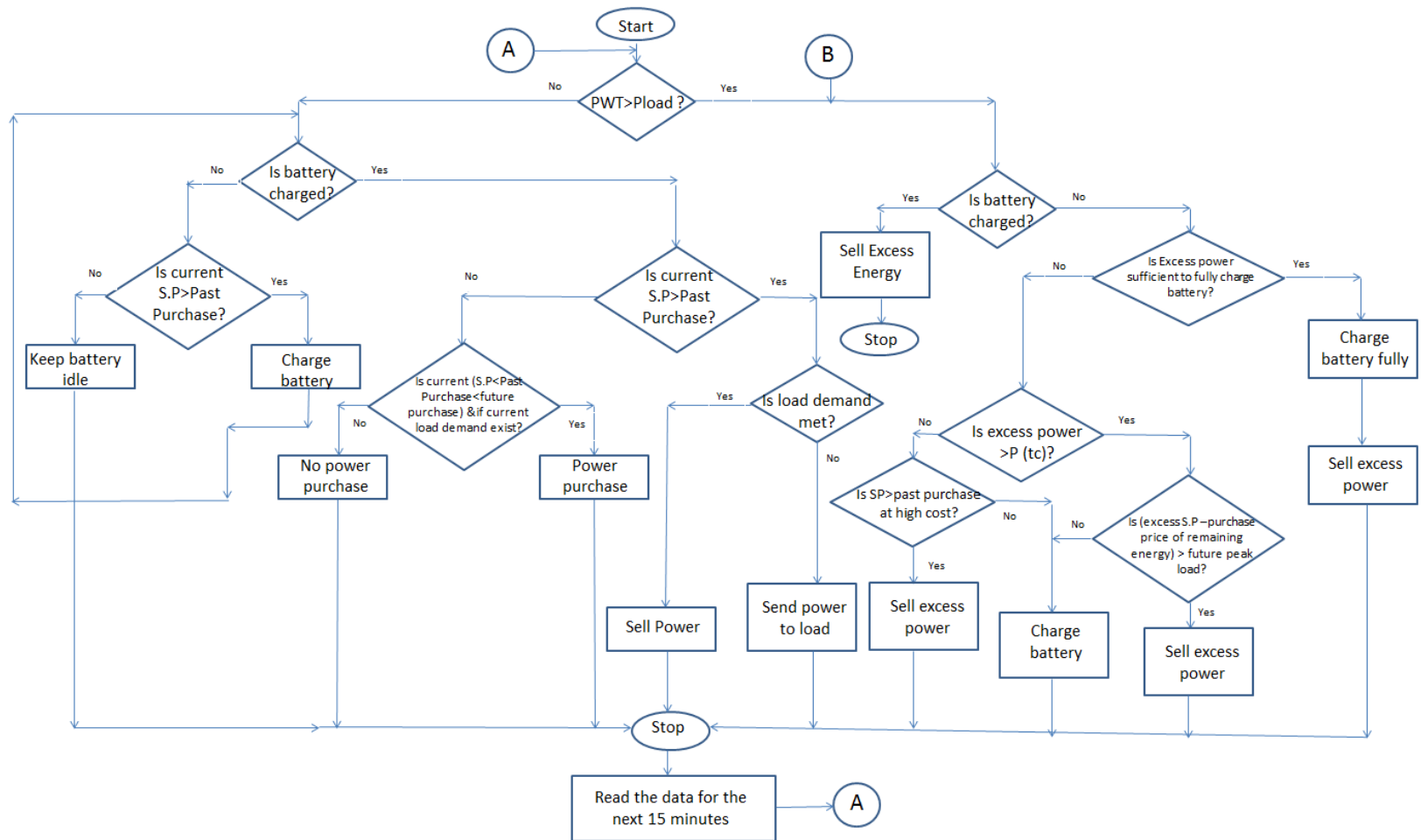


Figure 181. General Algorithm for Power Supply Greater or Lesser Than the Load

The flowchart below considers the possible cases when the power supply from the renewable source is less and unable to meet the load. The various battery charge states are considered and the decisions are made based on the differential cost of energy required to purchase or sell power.

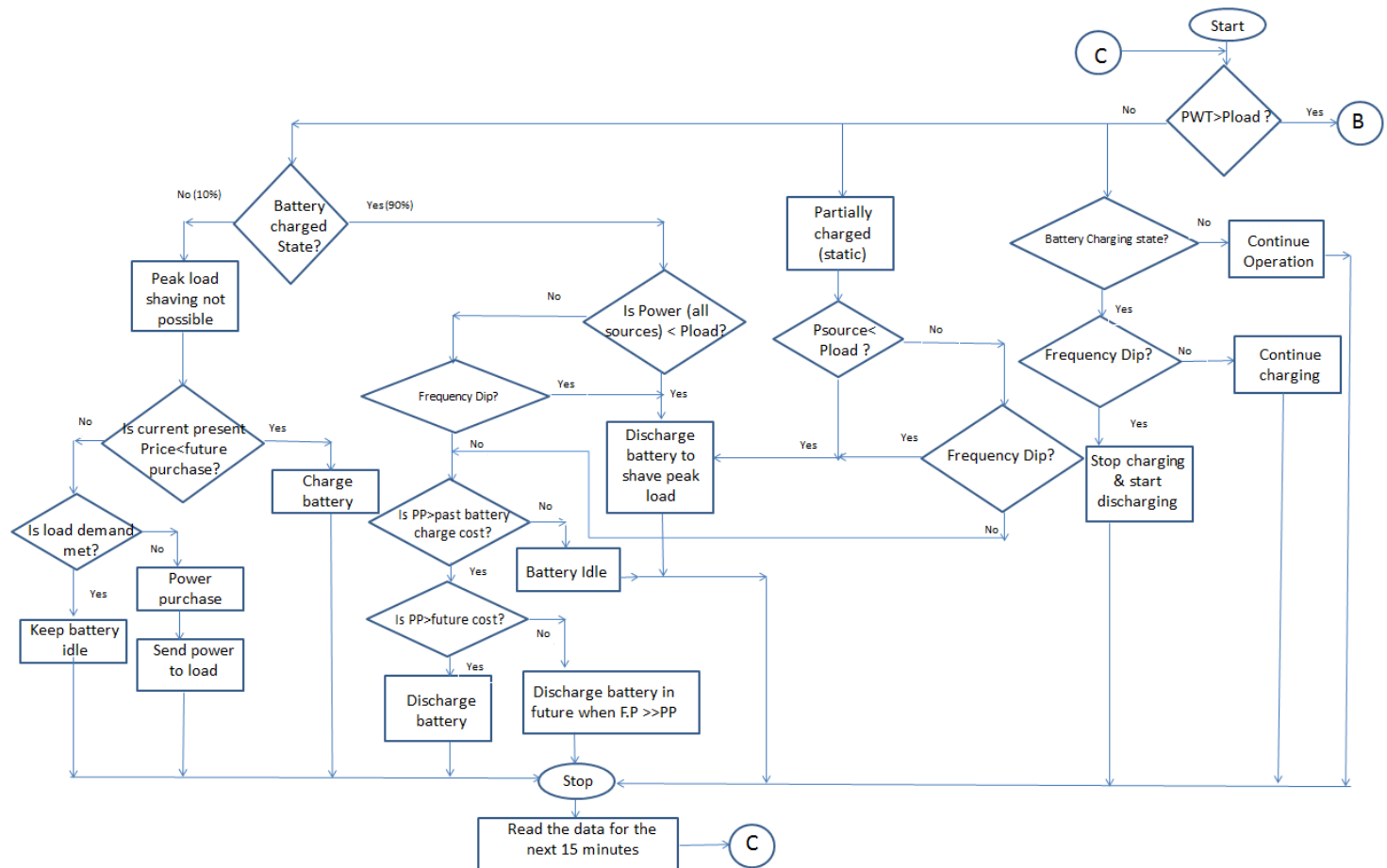


Figure 182. Algorithm with its Emphasis When Power Supply Lesser than the Load

8.5.4.2 Analysis and Simulation Results

The above flowcharts are explained as follows:

Assumptions:

- 1) Battery capacity is 1 MWh
- 2) Wind capacity is 1.7 MW
- 3) Maximum load is considered to be between 2 MW to 3 MW
- 4) Charge the battery with constant charge rate, i.e. constant C rate
- 5) If a battery is charged, it is charged to no more than 90% SOC
- 6) If a battery is discharged, it is discharged to at least 20% SOC
- 7) Program reads the data every 15 minutes from the Excel files

The Python program reads the data from the Excel files and presents the recommended battery action in a graphical user interface as shown below. Some of the conditions and cases from the flowcharts above are explained as follows:

Case 1:

If power from the wind turbine is greater than the load demand, and the battery is in discharged state, then the condition “If excess power is sufficient to charge the battery” is checked. Also “If Excess energy can be sold at higher prices in future than the present purchase price of energy at peak load or at high price”, the recommended action is to “Sell excess power. Charge battery in future”.

Battery Recommended Action

Time:	2013-12-01 00:00:00						
Battery Production (kW):	1000	Past LMP Price:	60	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	1700	Current LMP Price:	70	Battery Charge(0-100%):	9	Frequency Indicator:	Frequency Nominal
Full Load (kW):	1000	Future Projected Price:	55	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: Sell excess power. Charge battery in future.					
SKIP NEXT							
START							

Figure 183. Algorithm Output – Sell Excess Power & Charge Battery in Future

Case 2:

If power from the wind turbine is greater than the load demand, and the battery is in discharged state, then condition “If excess power is sufficient to charge the battery” is checked. Also “If excess energy cannot be sold at higher prices in future than the present purchase price of energy at peak load or at high price”, the recommended action is to “Charge battery with excess power”.

Battery Recommended Action							
Time:	2013-12-01 00:00:00						
Battery Production (kW):	1000	Past LMP Price:	60	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	1700	Current LMP Price:	55	Battery Charge(0-100%):	9	Frequency Indicator:	Frequency Nominal
Full Load (kW):	1000	Future Projected Price:	70	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: <div>Charge battery with excess power.</div>					
SKIP NEXT							
START							

Figure 184. Algorithm Output – Charge Battery

Case 3:

If power generated from the wind turbine is greater than the load demand, and the battery is in charged state, then the condition “If present selling price is greater than past energy to charge the battery” is checked. Then it checks “If the present price is greater than the future selling price”, and if true the recommended action is to “Sell excess energy. Discharge battery in present.” as shown in the figure below. If not, “Sell excess energy. Discharge battery in future” as shown in the second figure.

Battery Recommended Action							
Time:	2013-12-01 00:00:00						
Battery Production (kW):	1000	Past LMP Price:	55	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	1700	Current LMP Price:	70	Battery Charge(0-100%):	91	Frequency Indicator:	Frequency Nominal
Full Load (kW):	1000	Future Projected Price:	60	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: Sell Excess Energy. Discharge Battery in present.					
SKIP NEXT							
START							

Figure 185. Algorithm Output – Sell Excess Energy & Discharge Battery at Present

Battery Recommended Action							
Time:	2013-12-01 00:00:00						
Battery Production (kW):	1000	Past LMP Price:	55	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	1700	Current LMP Price:	60	Battery Charge(0-100%):	100	Frequency Indicator:	Frequency Nominal
Full Load (kW):	1000	Future Projected Price:	70	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: Sell Excess Energy. Discharge Battery in future.					
SKIP NEXT							
START							

Figure 186. Algorithm Output – Sell Excess Energy & Discharge Battery in Future

Case 4:

If power generated from the wind turbine is greater than the load demand, and the battery is in charged state, then it checks “If present selling price is less than past energy to charge the battery” with a recommended action to “Sell excess power and keep battery idle”.

Time:	2013-12-01 00:00:00						
Battery Production (kW):	1000	Past LMP Price:	60	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	1700	Current LMP Price:	55	Battery Charge(0-100%):	100	Frequency Indicator:	Frequency Nominal
Full Load (kW):	1000	Future Projected Price:	70	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: Sell Excess Energy, Keep Battery Idle					
SKIP NEXT							
START							

Figure 187. Algorithm Output – Sell Excess Power and Keep Battery Idle

Case 5:

If power generated from the wind turbine is less than the load demand, and the battery is in charged state, then the condition “If present purchase price is greater than past energy to charge the battery” is checked. Then “If the future price is less than the current purchase price” is true, the recommended action is to “Discharge battery to load. Sell excess power. Charge battery in future” as in the first figure below. If not, then “Discharge battery to load. Purchase power to supply load” as in the second figure.

Time:	2013-12-01 00:00:00						
Battery Production (kW):	8.05333333	Past LMP Price:	55.0	Peak Shaving Limit(kW):	4500	Peak Shaving Indicator:	Below Peak Shaving Limit
Alstom Wind Production (kW):	7.4125	Current LMP Price:	60	Battery Charge(0-100%):	100	Frequency Indicator:	Frequency Nominal
Full Load (kW):	3481.75	Future Projected Price:	45.0	Current Grid Frequency(Hz):	60	Battery Charge Status(y/n):	n
QUIT	NEXT DATA	Battery Action: Discharge Battery to Load, Sell Excess Power, Charge battery in future.					
SKIP NEXT							
START							

Figure 188. Algorithm Output – Discharge Battery to Load & Sell Excess Power

Battery Recommended Action					
Time:	2013-12-01 00:00:00				
Battery Production (kW):	8.05333333	Past LMP Price:	55	Peak Shaving Limit(kW):	4500
Alstom Wind Production (kW):	7.4125	Current LMP Price:	60	Battery Charge(0-100%):	100
Full Load (kW):	3481.75	Future Projected Price:	70	Current Grid Frequency(Hz):	60
<div>QUIT</div> <div>NEXT DATA</div>		<div>Battery Action:</div> <div>Discharge Battery to Load. Purchase Power to supply load.</div>			
<div>SKIP NEXT</div>					
<div>START</div>					

Figure 189. Algorithm Output – Discharge Battery in Future & Purchase Power for the Load

Case 6:

If power generated from the wind turbine is less than the load demand, and the battery is in charged state, then check the condition “If present purchase price is less than past energy to charge the battery”. If true, the recommended action is to “ Purchase power, send to the load. Keep battery idled”.

Battery Recommended Action					
Time:	2013-12-01 00:00:00				
Battery Production (kW):	8.05333333	Past LMP Price:	60.0	Peak Shaving Limit(kW):	4500
Alstom Wind Production (kW):	7.4125	Current LMP Price:	55	Battery Charge(0-100%):	100
Full Load (kW):	3481.75	Future Projected Price:	70	Current Grid Frequency(Hz):	60
<div>QUIT</div> <div>NEXT DATA</div>		<div>Battery Action:</div> <div>Purchase Power, send to the load. Keep Battery Idle.</div>			
<div>SKIP NEXT</div>					
<div>START</div>					

Figure 190. Algorithm Output – Purchase Power Now and Send it to the Load

Case 7:

If power generated from the wind turbine is less than the load demand, and the battery is in discharged state, then check the condition “If present purchase price is less than purchase price of energy to charge the battery in future.” If true, the recommended action is to “ Charge battery”.

Time: 2013-12-01 00:00:00			
Battery Production (kW): 8.05333333	Past LMP Price: 55.0	Peak Shaving Limit(kW): 4500	Peak Shaving Indicator: Below Peak Shaving Limit
Alstom Wind Production (kW): 7.4125	Current LMP Price: 55	Battery Charge(0-100%): 9	Frequency Indicator: Frequency Nominal
Full Load (kW): 3481.75	Future Projected Price: 70	Current Grid Frequency(Hz): 60	Battery Charge Status(y/n): n

QUIT

SKIP NEXT

START

NEXT DATA

Battery Action:

Charge Battery.

Figure 191. Algorithm Output – Charge Battery

As presented, the algorithm is useful to predict the maximum utilization of the battery in an optimized and effective way to maximize profit. The battery cost and operational cost were not included in this analysis (it is presumed that the battery was already deployed and the analysis looks at the optimal time for the battery action related to grid/load and wind production and not as a function of cost of the battery), but the operational condition of the battery is considered and the recommended battery action is decided by comparing with known past LMP values, present cost values and predicted future LMP data. The algorithm also analyzes the need for peak shaving and to alert the operator for any frequency variations.

8.5.4.3 Energy Storage Computational Tool Results

For identifying, quantifying and monetizing the benefits of grid-connected energy storage, DOE recommends using the Excel-based energy storage computational tool (ESCT) which was funded by DOE and developed by Navigant.

- The ESCT characterizes energy storage projects by identifying the storage technologies employed and identifying the applications pursued. These parameters are defined for two different modules, the asset characterization module and the data input module. Each requires various inputs, and these are outlined in the next two sub-sections.
- Based on this characterization, the tool identifies the economic, reliability and environmental benefits the storage project will yield.
- The ESCT guides the user in entering data required for calculating the monetary value of benefits and associated capital and O&M costs.
- The ESCT prepares graphs and tables that summarize the costs and benefits of the project to help illustrate the project's overall value.

Some flexibility is required in using the ESCT to perform the economic analysis since the true economic value of a BESS will be greatly dependent on various factors such as the owner, type of market, location, and functional applications. For example, utilities in a deregulated market typically cannot own generation assets like a BESS for the purpose of generating revenue, so they are limited to using a BESS

to perform functions such as voltage support or grid reliability for which they would receive the normal regulated rate of return. This is currently true in the ERCOT market for the large investor-owned utilities (IOUs), but not for smaller utilities owned by rural cooperatives or cities. In other markets, large utilities can and do own generation assets, and even in ERCOT non-utility generators could own a BESS. Therefore, it's difficult to quantify the value proposition for a BESS under a single set of conditions.

Also, it should be noted that the ESCT determines the potential applications (source of revenue/savings) based on a certain set of built-in conditions that can limit results. Also, the tool limits the number of applications to three, a primary and two secondary applications. While this may be sufficient in most cases, it would be beneficial to allow more because certain applications, such as demand charge management or renewables energy time-shift, are more appropriate at specific times of the year rather than throughout the entire year.

To most fairly represent the economic potential in the Texas market, the Team performed four different sets of analysis, as outlined below. These sets represent various types of utilities, in both regulated and deregulated markets and located in different regions of the State, using a BESS to address diverse needs, from improved electric service reliability to ancillary services. It should be noted that the ESCT estimates of benefits primarily accrue to the owner of the BESS and do not easily represent multiple functions carried out simultaneously by multiple market participants.

The representative configurations of ownership, location and functionality include:

To most fairly represent the potential, this section includes four different sets of analysis:

1. 1MW BESS on a distribution network owned by a utility in a deregulated market. In Texas, this would be an IOU
2. 1MW BESS on a distribution network owned by a utility in a regulated market (without demand charge management). In Texas, this would be a cooperative or a municipal utility
3. 1MW BESS on a distribution network owned by a utility in a regulated market (with demand charge management). In Texas, this would be a cooperative or a municipal utility
4. 1MW BESS on a transmission network owned by a generator in a regulated market

Additionally, since battery costs are expected to be reduced by 50% in the next three years or so, this section includes one final set of analysis that re-evaluates the economics of option #1 with a 1MW BESS that costs half of what it did in this project.

8.5.4.3.1 Value Proposition for the 1MW BESS Owned by an IOU in a Deregulated Market

Asset Characterization Module

The following table shows the parameters given as inputs to the ESCT for this option. They are based on a 1MW BESS deployed on a distribution network and owned by an IOU operating in a deregulated market.

Table 32. Parameters for Asset Characterization Module

Location of battery	Distribution
Market	Deregulated
Owner	Utility
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Average/Expected year over demand growth of electric system	2%
Expected life time	20 years
Average inflation rate	2%
Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$2,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Expected decommissioning & disposal cost	\$25,000
Primary application	Electric service reliability
Selected secondary application(s) *	T&D upgrade deferral, Voltage support

* For this option, the ESCT only offered these alternatives.

On simulating the module using the above mentioned inputs, the following figure shows the potential benefits.

Benefits

The benefits highlighted in blue below represent the benefits that your energy storage project will yield based on primary and secondary applications that are being pursued. For further explanation of a benefit click the button to the right of the benefit.

Arbitrage Revenue	Definition	Deferred Transmission Investments	Definition
Capacity Market Revenue	Definition	Deferred Distribution Investments	Definition
Ancillary Services Revenue	Definition	Deferred Generation Capacity Investments	Definition
Reduced Electricity Cost (Consumer)	Definition	Reduced Outages (Consumer)	Definition
Reduced Electricity Cost (Utility/Ratepayer)	Definition	Reduced Outages (Utility/Ratepayer)	Definition
Reduced Congestion Costs (Non-Utility Merchant)	Definition	Improved Power Quality	Definition
Reduced Congestion Costs (Utility/Ratepayer)	Definition	Reduced CO2 Emissions	Definition
Optimized Generator Operation (Non-Utility Merchant)	Definition	Reduced SOx Emissions	Definition
Optimized Generator Operation (Utility/Ratepayer)	Definition	Reduced NOx Emissions	Definition
Reduced Electricity Losses	Definition	Reduced PM Emissions	Definition

Previous Finish

Figure 192. Results of Asset Characterization Module

Data Input Module

The following table shows the default inputs available in the ESCT tool for data input module analysis.

Table 33. Review Benefit Calculation Inputs *

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Distribution Capacity Deferred	kVA	100%	100%	100%	50	50	50
Annual Fixed Charge Rate for Distribution Capital Investment	%	100%	100%	100%	11	11	11
Capital Cost of Deferred Distribution Capacity	\$/kVA	100%	100%	100%	\$ 420	\$ 420	\$ 420
Yearly O&M Costs of Deferred Distribution Capacity	\$/year	100%	100%	100%	\$ 10,000	\$ 10,000	\$ 10,000
Initial Year of Distribution Deferral	year	100%	100%	100%	2014	2014	2014
Final year of Distribution Deferral	year	100%	100%	100%	2016	2016	2016
Capital Cost of Conventional Voltage Support Solution	\$/kVAR	100%	100%	100%	\$ 18	\$ 18	\$ 18
Nameplate Reactive Power Capacity of Energy Storage	kVAR	100%	100%	100%	1,000	1,000	1,000
Annual Fixed Charge Rate for Voltage Support Capital Investment	%	100%	100%	100%	11	11	11
Capital Cost of Conventional Electric Service Reliability Solution	\$/kW	100%	100%	100%	\$ 1,000	\$ 1,000	\$ 1,000
Annual Fixed Charge Rate for Electric Service Reliability Capital Investment	%	100%	100%	100%	11	11	11

* The \$1,000/kW for capital cost of conventional electric service reliability is based on average 2013 cost estimates for utility-scale electricity generating plants defined in a report from the Energy Information Administration at http://www.eia.gov/forecasts/capitalcost/pdf/updated_capcost.pdf.

Computational Module with Reference Case and Sensitivity Case Analyses

The computational module generates results based on reference case analysis and sensitivity case analysis from the above mentioned inputs.

The following table shows the cumulative benefit and cost table, with the total cost of deployment and total net benefit shown on the bottom right of the table. As you'll note in these results, the ESCT estimated \$34,300 in deferred distribution investments, \$23,500 in improved power quality savings, and \$1,308,600 in reduced outages. This total gross benefit of \$1,366,400 results in a \$1,593,500 loss (considering the cost of the BESS, annual O&M expenses, etc.) for this option.

Table 34. Cumulative Benefit and Cost Table

Cumulative Benefit and Cost Table				
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+ Primary and Secondary Benefits - Total Present Value over the Deployment Period	= Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -	\$ -	\$ -
	Capacity Market Revenue	\$ -	\$ -	\$ -
	Ancillary Services Revenue	\$ -	\$ -	\$ -
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -	\$ -	\$ -
	Optimized Generator Operation	\$ -	\$ -	\$ -
	Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
T&D Capital Savings	Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
	Deferred Transmission Investments	\$ -	\$ -	\$ -
	Deferred Distribution Investments	\$ -	\$ 34,300	\$ 34,300.00
Energy Efficiency	Reduced Electricity Losses	\$ -	\$ -	\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Interruptions	Reduced Outages (Consumer)	\$ -	\$ -	\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -	\$ 1,308,600	\$ 1,308,600.00
Power Quality	Improved Power Quality	\$ -	\$ 23,500	\$ 23,500.00
	Reduced CO2 Emissions	\$ -	\$ -	\$ -
	Reduced SOx Emissions	\$ -	\$ -	\$ -
	Reduced NOx Emissions	\$ -	\$ -	\$ -
	Reduced PM Emissions	\$ -	\$ -	\$ -
Total Gross Benefit		\$ -	\$ 1,366,400	\$ 1,366,400.00
Costs				
Capital Cost of Deployment (fixed charge rate)				\$ 2,406,800.00
Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to decommissioning and disposal costs)				\$ 546,900.00
Total Annual Cost of Deployment				\$ 2,959,900.00
Total Net Benefit				\$ (1,593,500.00)

The following figures show the results of the sensitivity case analysis which, as shown in the first figure, uses a range from 80% (low) to 120% (high) of the reference case values.

Table 35. Sensitivity Case Analysis Settings

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Distribution Capacity Deferred	kVA	80%	100%	120%	40	50	60
Annual Fixed Charge Rate for Distribution Capital Investment	%	80%	100%	120%	9	11	13
Capital Cost of Deferred Distribution Capacity	\$/kVA	80%	100%	120%	\$ 336	\$ 420	\$ 504
Yearly O&M Costs of Deferred Distribution Capacity	\$/year	80%	100%	120%	\$ 8,000	\$ 10,000	\$ 12,000
Initial Year of Distribution Deferral	year	80%	100%	120%	1611.2	2014	2416.8
Final year of Distribution Deferral	year	80%	100%	120%	1612.8	2016	2419.2
Capital Cost of Conventional Voltage Support Solution	\$/kVAR	80%	100%	120%	\$ 14	\$ 18	\$ 22
Nameplate Reactive Power Capacity of Energy Storage	kVAR	80%	100%	120%	800	1,000	1,200
Annual Fixed Charge Rate for Voltage Support Capital Investment	%	80%	100%	120%	9	11	13
Capital Cost of Conventional Electric Service Reliability Solution	\$/kW	80%	100%	120%	\$ 800	\$ 1,000	\$ 1,200
Annual Fixed Charge Rate for Electric Service Reliability Capital Investment	%	80%	100%	120%	9	11	13

Table 36. Results of Sensitivity Case Analysis

Cumulative Gross Benefits over the Deployment Period (Present Value)	Low Case	Reference Case	High Case
Arbitrage Revenue	\$ -	\$ -	\$ -
Capacity Market Revenue	\$ -	\$ -	\$ -
Ancillary Services Revenue	\$ -	\$ -	\$ -
Optimized Generator Operation (Non-Utility Merchant)	\$ -	\$ -	\$ -
Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Transmission Investments	\$ -	\$ -	\$ -
Deferred Distribution Investments	\$ -	\$ 34,300	\$ -
Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Reduced Outages (Consumer)	\$ -	\$ -	\$ -
Reduced Outages (Utility/Ratepayer)	\$ 837,900	\$ 1,308,600	\$ 1,884,300
Improved Power Quality	\$ 12,000	\$ 23,500	\$ 34,000
Reduced CO2 Emissions	\$ -	\$ -	\$ -
Reduced SOx Emissions	\$ -	\$ -	\$ -
Reduced NOx Emissions	\$ -	\$ -	\$ -
Reduced PM Emissions	\$ -	\$ -	\$ -
Total Benefit	\$ 849,900	\$ 1,366,400	\$ 1,918,300

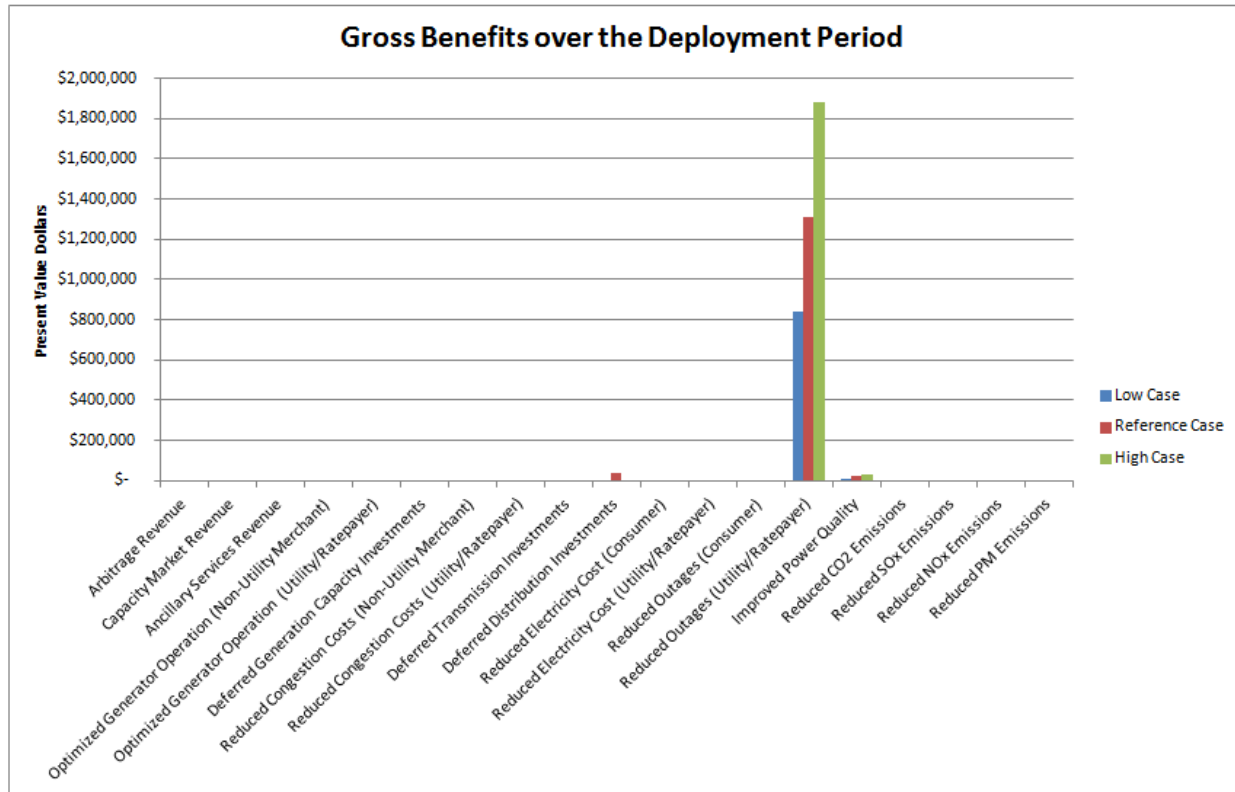


Figure 193. Graphic Results of Sensitivity Case Analysis

8.5.4.3.2 Value Proposition for the 1MW BESS Owned by a Utility in a Regulated Market

Asset Characterization Module

The following table shows the parameters given as inputs to the ESCT for this option. They are based on a 1MW BESS deployed on a distribution network and owned by a utility operating in a regulated market which means they can use the BESS to generate revenue. Although the ESCT does provide an application for demand charge management, this option below focuses more on renewables and wind integration support. The option presented in the next subsection will include demand charge management as an application.

Table 37. Parameters for Asset Characterization Module

Location of battery	Distribution
Market	Regulated
Owner	Utility
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Average/Expected year over demand growth of electric system	2%
Expected life time	20 years

Average inflation rate	2%
Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$2,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Expected decommissioning & disposal cost	\$25,000
Primary application	Renewables energy time-shift
Selected secondary application(s) *	Wind generation grid integration, voltage support

* For this option, the ESCT also offered two of the standard options, T&D upgrade deferral and electric service reliability, both of which were used in option #1 so the benefits of choosing those other options could be manually calculated.

On simulating the module using the above mentioned inputs, the following figure shows the potential benefits.

The benefits highlighted in blue below represent the benefits that your energy storage project will yield based on primary and secondary applications that are being pursued. For further explanation of a benefit click the button to the right of the benefit.

Arbitrage Revenue	Definition	Deferred Transmission Investments	Definition
Capacity Market Revenue	Definition	Deferred Distribution Investments	Definition
Ancillary Services Revenue	Definition	Deferred Generation Capacity Investments	Definition
Reduced Electricity Cost (Consumer)	Definition	Reduced Outages (Consumer)	Definition
Reduced Electricity Cost (Utility/Ratepayer)	Definition	Reduced Outages (Utility/Ratepayer)	Definition
Reduced Congestion Costs (Non-Utility Merchant)	Definition	Improved Power Quality	Definition
Reduced Congestion Costs (Utility/Ratepayer)	Definition	Reduced CO2 Emissions	Definition
Optimized Generator Operation (Non-Utility Merchant)	Definition	Reduced SOx Emissions	Definition
Optimized Generator Operation (Utility/Ratepayer)	Definition	Reduced NOx Emissions	Definition
Reduced Electricity Losses	Definition	Reduced PM Emissions	Definition

Previous Finish

Figure 194. Results of Asset Characterization Module

Data Input Module

The following table shows the inputs entered through the ESCT tool for data input module analysis.

Table 38. Review Benefit Calculation Inputs

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Average Variable Peak Generation Costs	\$/MWh	100%	100%	100%	\$ 70	\$ 70	\$ 70
CO2 Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
SOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	5.800	5.800	5.800
NOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.60	2.60	2.60
PM Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
Value of CO2	\$/ton	100%	100%	100%	\$ 20	\$ 20	\$ 20
Value of SOx	\$/ton	100%	100%	100%	\$ 520	\$ 520	\$ 520
Value of NOx	\$/ton	100%	100%	100%	\$ 3,000	\$ 3,000	\$ 3,000
Value of PM	\$/ton	100%	100%	100%	\$ 36,000	\$ 36,000	\$ 36,000
Average Variable Renewable Generation Costs	\$/MW	100%	100%	100%	\$ 2	\$ 2	\$ 2
Total Renewable Energy Discharged for Arbitrage	MWh	100%	100%	100%	782	782	782
Total Renewable Energy Discharged for Energy Time-Shift	MWh	100%	100%	100%	782	782	782
Capital Cost of Conventional Voltage Support Solution	\$/kVAR	100%	100%	100%	\$ 18	\$ 18	\$ 18
Nameplate Reactive Power Capacity of Energy Storage	kVAR	100%	100%	100%	1,000	1,000	1,000
Annual Fixed Charge Rate for Voltage Support Capital Investment	%	100%	100%	100%	11	11	11

Computational Module with Reference Case and Sensitivity Case Analyses

The computational module generates results based on reference case analysis and sensitivity case analysis from the above mentioned inputs.

The following table shows the cumulative benefit and cost table, based on the reference case analysis, with the total cost of deployment and total net benefit shown on the bottom right of the table. As you'll note in these results, for the 20-year period, the ESCT estimated \$632,800 in reduced electricity costs, \$23,500 in improved power quality savings, and \$546,600 in emissions savings. This total gross benefit of \$1,202,900 results in a \$1,757,000 loss (considering the cost of the BESS, annual O&M expenses, etc.) for this option.

Table 39. Cumulative Benefit and Cost Table

Cumulative Benefit and Cost Table				
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+ Primary and Secondary Benefits - Total Present Value over the Deployment Period	= Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -	\$ -	\$ -
	Capacity Market Revenue	\$ -	\$ -	\$ -
	Ancillary Services Revenue	\$ -	\$ -	\$ -
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -	\$ -	\$ -
	Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
	Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
T&D Capital Savings	Deferred Transmission Investments	\$ -	\$ -	\$ -
	Deferred Distribution Investments	\$ -	\$ -	\$ -
Energy Efficiency	Reduced Electricity Losses	\$ -	\$ -	\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ 632,800	\$ 632,800.00
Power Interruptions	Reduced Outages (Consumer)	\$ -	\$ -	\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Quality	Improved Power Quality	\$ -	\$ 23,500	\$ 23,500.00
Air Emissions	Reduced CO2 Emissions	\$ -	\$ 127,500	\$ 127,500.00
	Reduced SOx Emissions	\$ -	\$ 13,900	\$ 13,900.00
	Reduced NOx Emissions	\$ -	\$ 36,200	\$ 36,200.00
	Reduced PM Emissions	\$ -	\$ 369,000	\$ 369,000.00
	Total Gross Benefit	\$ -	\$ 1,202,900	\$ 1,202,900.00
Costs				
	Capital Cost of Deployment (fixed charge rate)			\$ 2,406,800.00
	Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to loss-			\$ 546,900.00
	Decommissioning and Disposal Costs			\$ 6,300.00
	Total Annual Cost of Deployment			\$ 2,959,900.00
	Total Net Benefit			\$ (1,757,000.00)

The following figures show the results of the sensitivity case analysis which, as shown in the first figure, uses a range from 80% (low) to 120% (high) of the reference case values.

Table 40. Sensitivity Case Analysis Settings

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Average Variable Peak Generation Costs	\$/MWh	80%	100%	120%	\$ 56	\$ 70	\$ 84
CO2 Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	1,098	1,372	1,646
SOx Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	4,640	5,800	6,960
NOx Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	2.08	2.60	3.12
PM Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	1.764	2.205	2.646
Value of CO2	\$/ton	80%	100%	120%	\$ 16	\$ 20	\$ 24
Value of SOx	\$/ton	80%	100%	120%	\$ 416	\$ 520	\$ 624
Value of NOx	\$/ton	80%	100%	120%	\$ 2,400	\$ 3,000	\$ 3,600
Value of PM	\$/ton	80%	100%	120%	\$ 28,800	\$ 36,000	\$ 43,200
Average Variable Renewable Generation Costs	\$/MW	80%	100%	120%	\$ 2	\$ 2	\$ 2
Total Renewable Energy Discharged for Arbitrage	MWh	80%	100%	120%	626	782	938
Total Renewable Energy Discharged for Energy Time-Shift	MWh	80%	100%	120%	626	782	938
Capital Cost of Conventional Voltage Support Solution	\$/kVAR	80%	100%	120%	\$ 14	\$ 18	\$ 22
Nameplate Reactive Power Capacity of Energy Storage	kVAR	80%	100%	120%	800	1,000	1,200
Annual Fixed Charge Rate for Voltage Support Capital Investment	%	80%	100%	120%	9	11	13

Table 41. Sensitivity Case Analysis Results

Cumulative Gross Benefits over the Deployment Period (Present Value)		Low Case	Reference Case	High Case
Arbitrage Revenue	\$ -	\$ -	\$ -	\$ -
Capacity Market Revenue	\$ -	\$ -	\$ -	\$ -
Ancillary Services Revenue	\$ -	\$ -	\$ -	\$ -
Optimized Generator Operation (Non-Utility Merchant)	\$ -	\$ -	\$ -	\$ -
Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -	\$ -
Deferred Generation Capacity Investments	\$ -	\$ -	\$ -	\$ -
Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -	\$ -
Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -	\$ -
Deferred Transmission Investments	\$ -	\$ -	\$ -	\$ -
Deferred Distribution Investments	\$ -	\$ -	\$ -	\$ -
Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -	\$ -
Reduced Electricity Cost (Utility/Ratepayer)	\$ 404,700	\$ 632,800	\$ 911,000	
Reduced Outages (Consumer)	\$ -	\$ -	\$ -	\$ -
Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -	\$ -
Improved Power Quality	\$ 12,000	\$ 23,500	\$ 34,000	
Reduced CO2 Emissions	\$ 65,500	\$ 127,500	\$ 220,600	
Reduced SOx Emissions	\$ 7,200	\$ 13,900	\$ 24,200	
Reduced NOx Emissions	\$ 18,900	\$ 36,200	\$ 62,700	
Reduced PM Emissions	\$ 189,100	\$ 369,000	\$ 637,900	
Total Benefit	\$ 697,400	\$ 1,202,900	\$ 1,890,400	

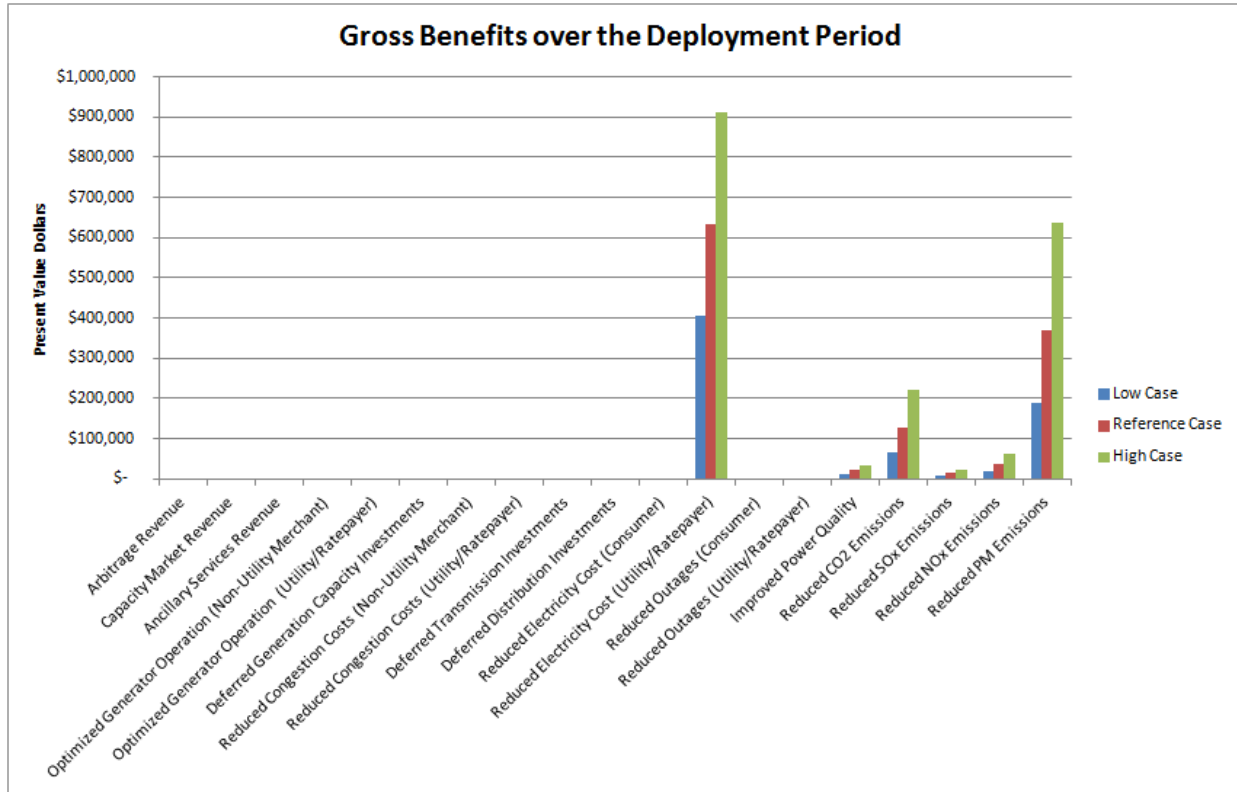


Figure 195. Graphic Representation of Sensitivity Case Analysis

8.5.4.3.3 Value Proposition for the 1MW BESS Owned by a Utility in a Regulated Market to Offset Demand Charges

Asset Characterization Module

The following table shows the parameters given as inputs to the ESCT for this option. They are based on a 1MW BESS deployed on a distribution network and owned by a utility operating in a regulated market which means they can use the BESS to generate revenue. For this option, the primary application is demand charge management.

Table 42. Parameters for Asset Characterization Module

Location of battery	Distribution
Market	Regulated
Owner	Utility (End User)
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Average/Expected year over demand growth of electric system	2%
Expected life time	20 years

Average inflation rate	2%
Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$2,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Expected decommissioning & disposal cost	\$25,000
Primary application	Demand charge management
Selected secondary application(s) *	Renewables energy time-shift, electric service power quality

* For this option, the ESCT also offered TOU energy cost management and electric service reliability, both of which were used in option #1 so their benefits could be considered.

On simulating the module using the above mentioned inputs, the following figure shows the potential benefits.

The benefits highlighted in blue below represent the benefits that your energy storage project will yield based on primary and secondary applications that are being pursued. For further explanation of a benefit click the button to the right of the benefit.

Arbitrage Revenue	Definition	Deferred Transmission Investments	Definition
Capacity Market Revenue	Definition	Deferred Distribution Investments	Definition
Ancillary Services Revenue	Definition	Deferred Generation Capacity Investments	Definition
Reduced Electricity Cost (Consumer)	Definition	Reduced Outages (Consumer)	Definition
Reduced Electricity Cost (Utility/Ratepayer)	Definition	Reduced Outages (Utility/Ratepayer)	Definition
Reduced Congestion Costs (Non-Utility Merchant)	Definition	Improved Power Quality	Definition
Reduced Congestion Costs (Utility/Ratepayer)	Definition	Reduced CO2 Emissions	Definition
Optimized Generator Operation (Non-Utility Merchant)	Definition	Reduced SOx Emissions	Definition
Optimized Generator Operation (Utility/Ratepayer)	Definition	Reduced NOx Emissions	Definition
Reduced Electricity Losses	Definition	Reduced PM Emissions	Definition

Previous Finish

Figure 196. Results of Asset Characterization Module

Data Input Module

The following table shows the inputs entered through the ESCT tool for data input module analysis.

Table 43. Review Benefit Calculation Inputs

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Average On-Peak Price of Electricity	\$/MWh	100%	100%	100%	\$ 100	\$ 100	\$ 100
CO2 Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
SOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	5.800	5.800	5.800
NOx Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.60	2.60	2.60
PM Emissions Factor for Generation on the Margin	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
Value of CO2	\$/ton	100%	100%	100%	\$ 20	\$ 20	\$ 20
Value of SOx	\$/ton	100%	100%	100%	\$ 520	\$ 520	\$ 520
Value of NOx	\$/ton	100%	100%	100%	\$ 3,000	\$ 3,000	\$ 3,000
Value of PM	\$/ton	100%	100%	100%	\$ 36,000	\$ 36,000	\$ 36,000
Average Monthly Load Reduction from ES at Peak	MW	100%	100%	100%	1.0	1.0	1.0
Average Monthly Load Reduction from ES at Partial-Peak	MW	100%	100%	100%	1.0	1.0	1.0
Average Monthly Load Reduction from ES at Off-Peak	MW	100%	100%	100%	1.0	1.0	1.0
Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 11,590	\$ 11,590	\$ 11,590
Partial-Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 2,650	\$ 2,650	\$ 2,650
Off-Peak Demand Charge	\$/MW-mc	100%	100%	100%	\$ 6,890	\$ 6,890	\$ 6,890
Number of Months that Demand Charges are Avoided	#	100%	100%	100%	5	5	5
Typical Number of Power Quality Events per Year	#	100%	100%	100%	10	10	10
Value of Avoided Power Quality Events	\$/kw peak	100%	100%	100%	\$ 5.0	\$ 5.0	\$ 5.0
Customer Peak Load	kW	100%	100%	100%	1,000	1,000	1,000
Total Renewable Energy Discharged for Arbitrage	MWh	100%	100%	100%	782	782	782
Total Renewable Energy Discharged for Energy Time-Shift	MWh	100%	100%	100%	782	782	782

Computational Module with Reference Case and Sensitivity Case Analyses

The computational module generates results based on reference case analysis and sensitivity case analysis from the above mentioned inputs.

The following table shows the cumulative benefit and cost table, based on the reference case analysis, with the total cost of deployment and total net benefit shown on the bottom right of the table. As you'll note in these results, for the 20-year period, the ESCT estimated \$2,309,600 in reduced electricity costs (mostly from reduced demand charges), \$594,900 in improved power quality savings (includes voltage and frequency support), and \$546,600 in emissions savings. This total gross benefit of \$3,451,000 results in a return on investment of \$491,200 (after the cost of the BESS, annual O&M expenses, etc.) for this option.

Table 44. Cumulative Benefit and Cost Table

	Cumulative Benefit and Cost Table					
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+	Primary and Secondary Benefits - Total Present Value over the Deployment Period	=	Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -		\$ -		\$ -
	Capacity Market Revenue	\$ -		\$ -		\$ -
	Ancillary Services Revenue	\$ -		\$ -		\$ -
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -		\$ -		\$ -
	Optimized Generator Operation (Utility/Ratepayer)	\$ -		\$ -		\$ -
	Deferred Generation Capacity Investments	\$ -		\$ -		\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -		\$ -		\$ -
	Reduced Congestion Costs (Utility/Ratepayer)	\$ -		\$ -		\$ -
T&D Capital Savings	Deferred Transmission Investments	\$ -		\$ -		\$ -
	Deferred Distribution Investments	\$ -		\$ -		\$ -
Energy Efficiency	Reduced Electricity Losses	\$ -		\$ -		\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -		2,309,600		\$ 2,309,600.00
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -		\$ -		\$ -
Power Interruptions	Reduced Outages (Consumer)	\$ -		\$ -		\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -		\$ -		\$ -
Power Quality	Improved Power Quality	\$ -		594,900		\$ 594,900.00
Air Emissions	Reduced CO2 Emissions	\$ -		127,500		\$ 127,500.00
	Reduced SOx Emissions	\$ -		13,900		\$ 13,900.00
	Reduced NOx Emissions	\$ -		36,200		\$ 36,200.00
	Reduced PM Emissions	\$ -		369,000		\$ 369,000.00
	Total Gross Benefit	\$ -		\$ 3,451,100		\$ 3,451,100.00
	Costs					
	Capital Cost of Deployment (fixed charge rate)					\$ 2,406,800.00
	Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to loss-					\$ 546,900.00
	Decommissioning and Disposal Costs					\$ 6,300.00
	Total Annual Cost of Deployment					\$ 2,959,900.00
	Total Net Benefit					\$ 491,200.00

The following figures show the results of the sensitivity case analysis which, as shown in the first figure, uses a range from 80% (low) to 120% (high) of the reference case values.

Table 45. Sensitivity Case Analysis Settings

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Average On-Peak Price of Electricity	\$/MWh	80%	100%	120%	\$ 80	\$ 100	\$ 120
CO2 Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	1,098	1,372	1,646
SOx Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	4,640	5,800	6,960
NOx Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	2.08	2.60	3.12
PM Emissions Factor for Generation on the Margin	lbs/MWh	80%	100%	120%	1.764	2.205	2.646
Value of CO2	\$/ton	80%	100%	120%	\$ 16	\$ 20	\$ 24
Value of SOx	\$/ton	80%	100%	120%	\$ 416	\$ 520	\$ 624
Value of NOx	\$/ton	80%	100%	120%	\$ 2,400	\$ 3,000	\$ 3,600
Value of PM	\$/ton	80%	100%	120%	\$ 28,800	\$ 36,000	\$ 43,200
Average Monthly Load Reduction from ES at Peak	MW	80%	100%	120%	0.8	1.0	1.2
Average Monthly Load Reduction from ES at Partial-Peak	MW	80%	100%	120%	0.8	1.0	1.2
Average Monthly Load Reduction from ES at Off-Peak	MW	80%	100%	120%	0.8	1.0	1.2
Peak Demand Charge	\$/MW-mo	80%	100%	120%	\$ 9,272	\$ 11,590	\$ 13,908
Partial-Peak Demand Charge	\$/MW-mo	80%	100%	120%	\$ 2,120	\$ 2,650	\$ 3,180
Off-Peak Demand Charge	\$/MW-mo	80%	100%	120%	\$ 5,512	\$ 6,890	\$ 8,268
Number of Months that Demand Charges are Avoided	#	80%	100%	120%	4	5	6
Typical Number of Power Quality Events per Year	#	80%	100%	120%	8	10	12
Value of Avoided Power Quality Events	\$/kw peak	80%	100%	120%	\$ 4.0	\$ 5.0	\$ 6.0
Customer Peak Load	kW	80%	100%	120%	800	1,000	1,200
Total Renewable Energy Discharged for Arbitrage	MWh	80%	100%	120%	626	782	938
Total Renewable Energy Discharged for Energy Time-Shift	MWh	80%	100%	120%	626	782	938

Table 46. Sensitivity Case Analysis Results

Cumulative Gross Benefits over the Deployment Period (Present Value)	Low Case	Reference Case	High Case
Arbitrage Revenue	\$ -	\$ -	\$ -
Capacity Market Revenue	\$ -	\$ -	\$ -
Ancillary Services Revenue	\$ -	\$ -	\$ -
Optimized Generator Operation (Non-Utility Merchant)	\$ -	\$ -	\$ -
Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Transmission Investments	\$ -	\$ -	\$ -
Deferred Distribution Investments	\$ -	\$ -	\$ -
Reduced Electricity Cost (Consumer)	\$ 1,372,100	\$ 2,309,600	\$ 3,325,700
Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Reduced Outages (Consumer)	\$ -	\$ -	\$ -
Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -
Improved Power Quality	\$ 331,900	\$ 594,900	\$ 856,600
Reduced CO2 Emissions	\$ 65,500	\$ 127,500	\$ 220,600
Reduced SOx Emissions	\$ 7,200	\$ 13,900	\$ 24,200
Reduced NOx Emissions	\$ 18,900	\$ 36,200	\$ 62,700
Reduced PM Emissions	\$ 189,100	\$ 369,000	\$ 637,900
Total Benefit	\$ 1,984,700	\$ 3,451,100	\$ 5,127,700

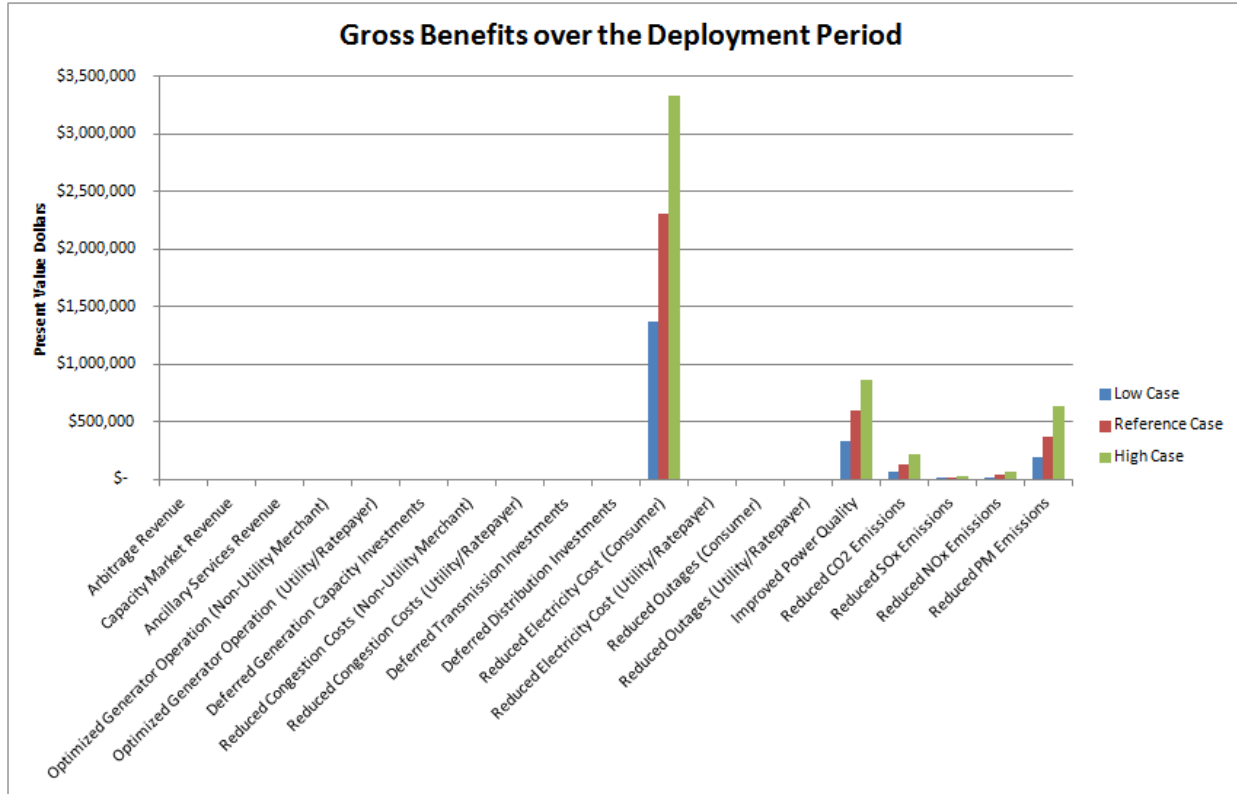


Figure 197. Graphic Representation of Sensitivity Case Analysis

8.5.4.3.4 Value Proposition for the 1MW BESS Owned by a Generator in a Deregulated Market

Asset Characterization Module

The following table shows the parameters given as inputs to the ESCT for this option. They are based on a 1MW BESS deployed on a distribution network and owned by a utility operating in a regulated market which means they can use the BESS to generate revenue.

Table 47. Parameters for Asset Characterization Module

Location of battery	Transmission
Market	Deregulated
Owner	Utility (End User)
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Average/Expected year over demand growth of electric system	2%
Expected life time	20 years
Average inflation rate	2%

Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$2,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Expected decommissioning & disposal cost	\$25,000
Primary application	Renewables capacity firming
Selected secondary application(s) *	Electric supply reserve capacity and voltage support

* For this option, the ESCT also offered renewables energy time-shift and wind generation grid integration, both of which were covered in other options.

On simulating the module using the above mentioned inputs, the following figure shows the potential benefits, which now include ancillary services.

The benefits highlighted in blue below represent the benefits that your energy storage project will yield based on primary and secondary applications that are being pursued. For further explanation of a benefit click the button to the right of the benefit.

Arbitrage Revenue	Definition	Deferred Transmission Investments	Definition
Capacity Market Revenue	Definition	Deferred Distribution Investments	Definition
Ancillary Services Revenue	Definition	Deferred Generation Capacity Investments	Definition
Reduced Electricity Cost (Consumer)	Definition	Reduced Outages (Consumer)	Definition
Reduced Electricity Cost (Utility/Ratepayer)	Definition	Reduced Outages (Utility/Ratepayer)	Definition
Reduced Congestion Costs (Non-Utility Merchant)	Definition	Improved Power Quality	Definition
Reduced Congestion Costs (Utility/Ratepayer)	Definition	Reduced CO2 Emissions	Definition
Optimized Generator Operation (Non-Utility Merchant)	Definition	Reduced SOx Emissions	Definition
Optimized Generator Operation (Utility/Ratepayer)	Definition	Reduced NOx Emissions	Definition
Reduced Electricity Losses	Definition	Reduced PM Emissions	Definition

Previous Finish

Figure 198. Results of Asset Characterization Module

Data Input Module

The following table shows the inputs entered through the ESCT tool for data input module analysis.

Table 48. Review Benefit Calculation Inputs

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
CO2 Emissions Factor for Base Generation	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
SOx Emissions Factor for Base Generation	lbs/MWh	100%	100%	100%	5.80	5.80	5.80
NOx Emissions Factor for Base Generation	lbs/MWh	100%	100%	100%	2.60	2.60	2.60
PM Emissions Factor for Base Generation	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
Value of CO2	\$/ton	100%	100%	100%	\$ 20	\$ 20	\$ 20
Value of SOx	\$/ton	100%	100%	100%	\$ 520	\$ 520	\$ 520
Value of NOx	\$/ton	100%	100%	100%	\$ 3,000	\$ 3,000	\$ 3,000
Value of PM	\$/ton	100%	100%	100%	\$ 36,000	\$ 36,000	\$ 36,000
Average Value of Capacity on the Market	\$/MW-h	100%	100%	100%	\$ 18	\$ 18	\$ 18
Total Spinning-Reserve Services Provided	MW-h	100%	100%	100%	2,000	2,000	2,000
Average Price for Spinning Reserve Services	\$/MW-h	100%	100%	100%	\$ 4.50	\$ 4.50	\$ 4.50
Total Non-Spinning-Reserve Services Provided	MW-h	100%	100%	100%	500	500	500
Average Price for Non-Spinning Reserve Services	\$/MW-h	100%	100%	100%	\$ 3.00	\$ 3.00	\$ 3.00
Total Backup Supply Services Provided	MW-h	100%	100%	100%	150	150	150
Average Price for Backup Supply Services	\$/MW-h	100%	100%	100%	\$ 0.40	\$ 0.40	\$ 0.40
Marginal Operating Costs of Conventional Generation at Partial Load	\$/MWh	100%	100%	100%	\$ 46.06	\$ 46.06	\$ 46.06
Marginal Operating Costs of Conventional Generation at Optimal Load	\$/MWh	100%	100%	100%	\$ 45.60	\$ 45.60	\$ 45.60
CO2 Emissions Factor of Generation at Partial Load	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
CO2 Emissions Factor of Generation at Optimal Load	lbs/MWh	100%	100%	100%	1,372	1,372	1,372
SOx Emissions Factor of Generation at Partial Load	lbs/MWh	100%	100%	100%	5.800	5.800	5.800
SOx Emissions Factor of Generation at Optimal Load	lbs/MWh	100%	100%	100%	5.800	5.800	5.800
NOx Emissions Factor of Generation at Partial Load	lbs/MWh	100%	100%	100%	2.600	2.600	2.600
NOx Emissions Factor of Generation at Optimal Load	lbs/MWh	100%	100%	100%	2.600	2.600	2.600
PM Emissions Factor of Generation at Partial Load	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
PM Emissions Factor of Generation at Optimal Load	lbs/MWh	100%	100%	100%	2.205	2.205	2.205
Annual Reactive Power Capacity Available for Voltage Support	MVAR-yr	100%	100%	100%	500.0	500.0	500.0
Annual Capacity Payment for Voltage Support	\$/MVAR-yr	100%	100%	100%	\$ 1,000	\$ 1,000	\$ 1,000
Effective Load Carrying Capacity of Renewable Post-Firming	%	100%	100%	100%	10	10	10
Effective Load Carrying Capacity of Renewable Pre-Firming	%	100%	100%	100%	10	10	10
Hours of Renewable Energy Capacity Sold	hr	100%	100%	100%	7,500.0	7,500.0	7,500.0
Nameplate Capacity of Renewable Resource	MW	100%	100%	100%	10	10	10
Capacity Factor of Renewable Resource	%	100%	100%	100%	12	12	12

Computational Module with Reference Case and Sensitivity Case Analyses

The computational module generates results based on reference case analysis and sensitivity case analysis from the above mentioned inputs.

The following table shows the cumulative benefit and cost table, based on the reference case analysis, with the total cost of deployment and total net benefit shown on the bottom right of the table. As you'll note in these results, for the 20-year period, the ESCT estimated \$736,100 in ancillary services revenue and \$11,800 in improved asset utilization. This total gross benefit of \$747,900 results in a loss of

\$2,212,000 (after the cost of the BESS, annual O&M expenses, etc.) for this option. However, this option was only run to capture the ancillary services potential, which could be added to other options.

Table 49. Cumulative Benefit and Cost Table

Cumulative Benefit and Cost Table				
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+ Primary and Secondary Benefits - Total Present Value over the Deployment Period	= Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -	\$ -	\$ -
	Capacity Market Revenue	\$ -	\$ -	\$ -
	Ancillary Services Revenue	\$ -	\$ 736,100	\$ 736,100.00
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -	\$ 11,800	\$ 11,800.00
	Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
	Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
T&D Capital Savings	Deferred Transmission Investments	\$ -	\$ -	\$ -
	Deferred Distribution Investments	\$ -	\$ -	\$ -
Energy Efficiency	Reduced Electricity Losses	\$ -	\$ -	\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Interruptions	Reduced Outages (Consumer)	\$ -	\$ -	\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Quality	Improved Power Quality	\$ -	\$ -	\$ -
Air Emissions	Reduced CO2 Emissions	\$ -	\$ -	\$ -
	Reduced SOx Emissions	\$ -	\$ -	\$ -
	Reduced NOx Emissions	\$ -	\$ -	\$ -
	Reduced PM Emissions	\$ -	\$ -	\$ -
	Total Gross Benefit	\$ -	\$ 747,900	\$ 747,900.00
Costs				
	Capital Cost of Deployment (fixed charge rate)			\$ 2,406,800.00
	Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to loss-			\$ 546,900.00
	Decommissioning and Disposal Costs			\$ 6,300.00
	Total Annual Cost of Deployment			\$ 2,959,900.00
	Total Net Benefit			\$ (2,212,000.00)

The following figures show the results of the sensitivity case analysis which, as shown in the first figure, uses a range from 80% (low) to 120% (high) of the reference case values.

Table 50. Sensitivity Case Analysis Settings

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
CO2 Emissions Factor for Base Generation	lbs/MWh	80%	100%	120%	1,098	1,372	1,646
SOx Emissions Factor for Base Generation	lbs/MWh	80%	100%	120%	4.64	5.80	6.96
NOx Emissions Factor for Base Generation	lbs/MWh	80%	100%	120%	2.08	2.60	3.12
PM Emissions Factor for Base Generation	lbs/MWh	80%	100%	120%	1.764	2.205	2.646
Value of CO2	\$/ton	80%	100%	120%	\$ 16	\$ 20	\$ 24
Value of SOx	\$/ton	80%	100%	120%	\$ 416	\$ 520	\$ 624
Value of NOx	\$/ton	80%	100%	120%	\$ 2,400	\$ 3,000	\$ 3,600
Value of PM	\$/ton	80%	100%	120%	\$ 28,800	\$ 36,000	\$ 43,200
Average Value of Capacity on the Market	\$/MW-h	80%	100%	120%	\$ 15	\$ 18	\$ 22
Total Spinning-Reserve Services Provided	MW-h	80%	100%	120%	1,600	2,000	2,400
Average Price for Spinning Reserve Services	\$/MW-h	80%	100%	120%	\$ 3.60	\$ 4.50	\$ 5.40
Total Non-Spinning-Reserve Services Provided	MW-h	80%	100%	120%	400	500	600
Average Price for Non-Spinning Reserve Services	\$/MW-h	80%	100%	120%	\$ 2.40	\$ 3.00	\$ 3.60
Total Backup Supply Services Provided	MW-h	80%	100%	120%	120	150	180
Average Price for Backup Supply Services	\$/MW-h	80%	100%	120%	\$ 0.32	\$ 0.40	\$ 0.48
Marginal Operating Costs of Conventional Generation at Partial Load	\$/MWh	80%	100%	120%	\$ 36.84	\$ 46.06	\$ 55.27
Marginal Operating Costs of Conventional Generation at Optimal Load	\$/MWh	80%	100%	120%	\$ 36.48	\$ 45.60	\$ 54.72
CO2 Emissions Factor of Generation at Partial Load	lbs/MWh	80%	100%	120%	1,098	1,372	1,646
CO2 Emissions Factor of Generation at Optimal Load	lbs/MWh	80%	100%	120%	1,098	1,372	1,646
SOx Emissions Factor of Generation at Partial Load	lbs/MWh	80%	100%	120%	4.640	5.800	6.960
SOx Emissions Factor of Generation at Optimal Load	lbs/MWh	80%	100%	120%	4.640	5.800	6.960
NOx Emissions Factor of Generation at Partial Load	lbs/MWh	80%	100%	120%	2.080	2.600	3.120
NOx Emissions Factor of Generation at Optimal Load	lbs/MWh	80%	100%	120%	2.080	2.600	3.120
PM Emissions Factor of Generation at Partial Load	lbs/MWh	80%	100%	120%	1.764	2.205	2.646
PM Emissions Factor of Generation at Optimal Load	lbs/MWh	80%	100%	120%	1.764	2.205	2.646
Annual Reactive Power Capacity Available for Voltage Support	MVAR-yr	80%	100%	120%	400.0	500.0	600.0
Annual Capacity Payment for Voltage Support	\$/MVAR-yr	80%	100%	120%	\$ 800	\$ 1,000	\$ 1,200
Effective Load Carrying Capacity of Renewable Post-Firming	%	80%	100%	120%	8	10	12
Effective Load Carrying Capacity of Renewable Pre-Firming	%	80%	100%	120%	8	10	12
Hours of Renewable Energy Capacity Sold	hr	80%	100%	120%	6,000.0	7,500.0	9,000.0
Nameplate Capacity of Renewable Resource	MW	80%	100%	120%	8	10	12
Capacity Factor of Renewable Resource	%	80%	100%	120%	10	12	14

Table 51. Sensitivity Case Analysis Results

Cumulative Gross Benefits over the Deployment Period (Present Value)	Low Case	Reference Case	High Case
Arbitrage Revenue	\$ -	\$ -	\$ -
Capacity Market Revenue	\$ -	\$ -	\$ -
Ancillary Services Revenue	\$ 471,000	\$ 736,100	\$ 1,059,800
Optimized Generator Operation (Non-Utility Merchant)	\$ 7,600	\$ 11,800	\$ 17,200
Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
Deferred Transmission Investments	\$ -	\$ -	\$ -
Deferred Distribution Investments	\$ -	\$ -	\$ -
Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Reduced Outages (Consumer)	\$ -	\$ -	\$ -
Reduced Outages (Utility/Ratepayer)	\$ -	\$ -	\$ -
Improved Power Quality	\$ -	\$ -	\$ -
Reduced CO2 Emissions	\$ -	\$ -	\$ -
Reduced SOx Emissions	\$ -	\$ -	\$ -
Reduced NOx Emissions	\$ -	\$ -	\$ -
Reduced PM Emissions	\$ -	\$ -	\$ -
Total Benefit	\$ 478,600	\$ 747,900	\$ 1,077,000

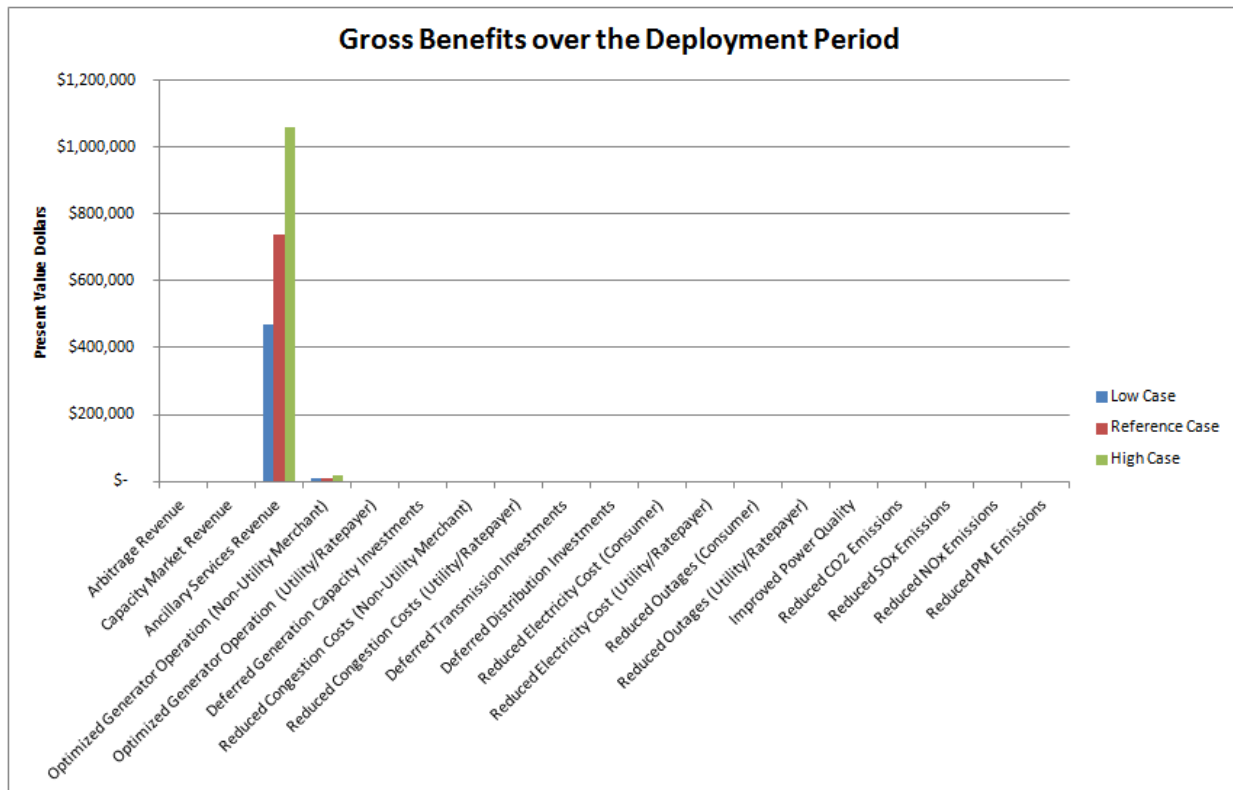


Figure 199. Graphic Representation of Sensitivity Case Analysis

8.5.4.3.5 Value Proposition for the 1MW BESS Owned by a Utility in a Deregulated Market – Reducing the BESS Cost by 50%

Asset Characterization Module

As shown earlier for option #1, the following table shows the parameters given as inputs to the ESCT for this option. They are based on a 1MW BESS deployed on a distribution network and owned by a utility operating in a deregulated market. In this option, the total installed cost of deployment has been cut in half, and is now \$1,000,000.

Table 52. Parameters for Asset Characterization Module

Location of battery	Distribution
Market	Deregulated
Owner	Utility
Type of Battery	Lithium-ion
Total name plate power output	1000kW
Total name plate energy storage capacity	1000kWh
Round trip efficiency	95%
Response time of BESS	0.005 seconds
Cycle life	4000 cycles
Average/Expected year over demand growth of electric system	2%
Expected life time	20 years
Average inflation rate	2%
Discount rate in net present value analysis	7%
Total Installed cost of deployment	\$1,000,000
Fixed charge rate to annualize cost of deployment	11%
Non energy O&M cost with replacement cost	\$50,000/year
Expected decommissioning & disposal cost	\$25,000
Primary application	Electric service reliability
Selected secondary application(s) *	T&D upgrade deferral, Voltage support

* For this option, the ESCT only offered these alternatives.

On simulating the module using the above mentioned inputs, the following figure shows the potential benefits.

Benefits

The benefits highlighted in blue below represent the benefits that your energy storage project will yield based on primary and secondary applications that are being pursued. For further explanation of a benefit click the button to the right of the benefit.

Arbitrage Revenue	Definition	Deferred Transmission Investments	Definition
Capacity Market Revenue	Definition	Deferred Distribution Investments	Definition
Ancillary Services Revenue	Definition	Deferred Generation Capacity Investments	Definition
Reduced Electricity Cost (Consumer)	Definition	Reduced Outages (Consumer)	Definition
Reduced Electricity Cost (Utility/Ratepayer)	Definition	Reduced Outages (Utility/Ratepayer)	Definition
Reduced Congestion Costs (Non-Utility Merchant)	Definition	Improved Power Quality	Definition
Reduced Congestion Costs (Utility/Ratepayer)	Definition	Reduced CO2 Emissions	Definition
Optimized Generator Operation (Non-Utility Merchant)	Definition	Reduced SOx Emissions	Definition
Optimized Generator Operation (Utility/Ratepayer)	Definition	Reduced NOx Emissions	Definition
Reduced Electricity Losses	Definition	Reduced PM Emissions	Definition

Previous Finish

Figure 200. Results of Asset Characterization Module

Data Input Module

The following table shows the inputs entered through the ESCT tool for data input module analysis.

Table 53. Review Benefit Calculation Inputs

Input Name	Unit	Select % using toggle			2014 Values		
		Low	Reference	High	Low	Reference	High
Distribution Capacity Deferred	kVA	100%	100%	100%	50	50	50
Annual Fixed Charge Rate for Distribution Capital Investment	%	100%	100%	100%	11	11	11
Capital Cost of Deferred Distribution Capacity	\$/kVA	100%	100%	100%	\$ 420	\$ 420	\$ 420
Yearly O&M Costs of Deferred Distribution Capacity	\$/year	100%	100%	100%	\$ 10,000	\$ 10,000	\$ 10,000
Initial Year of Distribution Deferral	year	100%	100%	100%	2014	2014	2014
Final year of Distribution Deferral	year	100%	100%	100%	2016	2016	2016
Capital Cost of Conventional Voltage Support Solution	\$/kVAR	100%	100%	100%	\$ 18	\$ 18	\$ 18
Nameplate Reactive Power Capacity of Energy Storage	kVAR	100%	100%	100%	1,000	1,000	1,000
Annual Fixed Charge Rate for Voltage Support Capital Investment	%	100%	100%	100%	11	11	11
Capital Cost of Conventional Electric Service Reliability Solution	\$/kW	100%	100%	100%	\$ 1,000	\$ 1,000	\$ 1,000
Annual Fixed Charge Rate for Electric Service Reliability Capital Investment	%	100%	100%	100%	11	11	11

Computational Module with Reference Case and Sensitivity Case Analyses

The computational module generates results based on reference case analysis and sensitivity case analysis from the above mentioned inputs.

The following table shows the cumulative benefit and cost table, based on the reference case analysis, with the total cost of deployment and total net benefit shown on the bottom right of the table. For the earlier option #1, this total gross benefit of \$1,366,400 resulted in a \$1,593,500 loss. In this option #5, with the BESS cost reduced to \$1,000,000, the capital cost of deployment was reduced, so the loss over the 20-year period is now significantly less at \$390,200.

Table 54. Cumulative Benefit and Cost Table

Cumulative Benefit and Cost Table				
	Benefits	Additional Benefits - Total Present Value over the Deployment Period	+ Primary and Secondary Benefits - Total Present Value over the Deployment Period	= Total Benefit - Present Value over the Deployment Period
Market Revenue	Arbitrage Revenue	\$ -	\$ -	\$ -
	Capacity Market Revenue	\$ -	\$ -	\$ -
	Ancillary Services Revenue	\$ -	\$ -	\$ -
Improved Asset Utilization	Optimized Generator Operation (Non-Utility)	\$ -	\$ -	\$ -
	Optimized Generator Operation (Utility/Ratepayer)	\$ -	\$ -	\$ -
	Deferred Generation Capacity Investments	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Non-Utility Merchant)	\$ -	\$ -	\$ -
	Reduced Congestion Costs (Utility/Ratepayer)	\$ -	\$ -	\$ -
T&D Capital Savings	Deferred Transmission Investments	\$ -	\$ -	\$ -
	Deferred Distribution Investments	\$ -	\$ 34,300	\$ 34,300.00
Energy Efficiency	Reduced Electricity Losses	\$ -	\$ -	\$ -
Electricity Cost Savings	Reduced Electricity Cost (Consumer)	\$ -	\$ -	\$ -
	Reduced Electricity Cost (Utility/Ratepayer)	\$ -	\$ -	\$ -
Power Interruptions	Reduced Outages (Consumer)	\$ -	\$ -	\$ -
	Reduced Outages (Utility/Ratepayer)	\$ -	\$ 1,308,600	\$ 1,308,600.00
Power Quality	Improved Power Quality	\$ -	\$ 23,500	\$ 23,500.00
Air Emissions	Reduced CO2 Emissions	\$ -	\$ -	\$ -
	Reduced SOx Emissions	\$ -	\$ -	\$ -
	Reduced NOx Emissions	\$ -	\$ -	\$ -
	Reduced PM Emissions	\$ -	\$ -	\$ -
	Total Gross Benefit	\$ -	\$ 1,366,400	\$ 1,366,400.00
Costs				
	Capital Cost of Deployment (fixed charge rate)			\$ 1,203,100.00
	Operating and maintenance costs not related to energy (labor for operation, plant maintenance, equipment wear leading to loss- Decommissioning and Disposal Costs)			\$ 546,900.00 \$ 6,300.00
	Total Annual Cost of Deployment			\$ 1,756,600.00
	Total Net Benefit			\$ (390,200.00)

If the market can reduce the cost of a BESS by 50% or so, then by adding other potential applications, such as demand charge reduction and ancillary services, it's conceivable that a BESS could be a break-even proposition or even generate net revenue over a 20-year period.

It should be noted that the net benefits calculated by this ESCT include considerable value for environmental benefits that cannot be directly captured by a utility making the investment today. The rationale of the tool is that energy storage systems can reduce system losses by reducing system peak demand which translates into a reduction in emissions if peak load losses are reduced by a significant degree compared to the slight increase in off-peak losses and/or if

peak generation units have lower or similar emissions factors to baseload units. Alternatively, by providing certain ancillary services, storage systems can enable conventional generation resources to be operated at more optimal conditions resulting in an emissions benefit. The calculation for this benefit assumes standard values for emission reductions based on estimates from literature. A future regulatory system may create an active emissions market allowing better estimates of these benefits.

8.5.5 Metrics and Benefits Analysis

This section covers the build and impact metrics for the BESS.

8.5.5.1 Build Metrics and Benefits

The base cost for development, testing and fielding of the 1MWh BESS was \$1,700,000. An additional \$295,933 was spent to prepare the site, implement monitoring and communications, and interconnect the BESS with the local grid. An additional \$40,000 was expended for a short-term vendor support contract. These costs do not include the wind turbines nor do they include the synchrophasor monitoring network. They also do not include costs incurred by SPEC in managing and monitoring the BESS.

8.5.5.2 Impact Metrics and Benefits

The impact metrics typically define quantitative energy or emission savings, while benefits may not be easily quantifiable. The benefits include:

- Optimized generator operation: Storage devices used will be providing energy during peak times and may enable conventional generators to operate closer to their optimal loading conditions thus improving overall generation system performance.
- Deferred generation capacity investments: To defer a transmission or distribution upgrade using energy storage, the unit is located downstream of strained assets and charged during off-peak times and discharged during peak times. This will effectively shift load away from the peak time of the system reducing the need for utility investments in generation capacity to meet peak load.
- Reduced electricity losses: To defer a transmission or distribution upgrade using energy storage, the unit is located downstream of strained assets and charged during off-peak times and discharged during peak times. This would effectively reduce the peak load on the transmission and distribution system, thereby reducing losses overall.

There are impact metrics associated with energy savings and reduced CO₂, SO_x, NO_x, and PM emissions.

- The BESS is used to shift peak electricity usage as a means of saving demand charges, but the actual amounts will vary depending on the actual peaks, the cost of electricity driving the

demand charges, how the demand charges are derived (highest versus rolling average), and the accuracy of forecasting peaks coupled with the ability of battery operators to have the system fully charged when needed.

- The reduced emissions use the following formulas:
 - CO₂: 0.00068956 tons/kWh
 - SO_x: 0.00263084 kg/kWh
 - NO_x: 0.00117934 kg/kWh
 - PM_{2.5}: 0.001 kg/kWh
- The savings over a 20-year period based on the ESCT results are as follows:
 - CO₂: \$127,500
 - SO_x: \$13,900
 - NO_x: \$36,200
 - PM_{2.5}: \$369,000

8.5.6 Stakeholder Feedback

To date, there are no documented observations from stakeholders on the impact of this project. ERCOT, SPP, GSEC and the CCET utility members have all shown an interest in energy storage, but as the testing of the BESS wasn't completed until late this year, the stakeholders have not yet had time to study the results of this analysis. Two TDSPs have deployed small community systems as a way of evaluating the potential for that use, and one TDSP even funded a study by the Brattle Group⁴⁰ to evaluate the economic feasibility of storage in Texas.

8.6 Conclusions

This section covers projections of demonstration and commercial scale system performance as well as lessons learned and best practices.

8.6.1 Projections of Demonstration and Commercial Scale System Performance

The BESS has proven that such a system can service a distribution utility and support multiple functions, individually or combined to optimize operational needs, reduce operational costs, or provide ancillary services that will generate return revenue to offset costs. The project demonstrated how effectively a 1MW BESS can be leveraged to level loads that are served by intermittent renewables, such as solar and wind. It proved how a BESS can reduce demand charges through peak shaving, while also providing invaluable frequency support and perhaps participation in ancillary services such as FRRS. Although a BESS of this scale is expensive, in the future utilities may be able to fund these resources through a rate

⁴⁰ The Value of Distributed Electricity Storage in Texas
http://www.brattle.com/system/news/pdfs/000/000/749/original/The_Value_of_Distributed_Electricity_Storage_in_Texas.pdf

case, and it is anticipated that the costs of energy storage devices will be halved in the next 3-5 years, making such investments much more attractive.

8.6.2 Lessons Learned and Best Practices

The following outlines some of the key lessons learned and best practices for fielding a BESS.

- From the time one begins drafting a request for proposal to the date of actual installation and operation of a system like the BESS can take as much as a year. The team began drafting its technical specification, and the actual request for proposal in October 2012. The solicitations were released in early November, and the proposals were evaluated and negotiated during December (this process typically takes 1-2 months longer). The factory acceptance test was performed in July 2013, and the BESS was deployed, installed and energized in August 2013.
- There are items, such as batteries and inverters, with long lead times, and even with proper planning, schedule delays can occur. Although some projects have experienced delays in procuring the necessary amount of batteries, this project did not have such an issue, but it did incur an unexpected one-month delay in receiving the requisite transformer.
- Financial stability of the battery system vendor is paramount given the lengthy development period. Typically, the vendor is integrating components, like the batteries, from other suppliers, and it's important to ensure that they all get paid. The team implemented a phased pay-out schedule tied to specific deliverables and, in each case, required verification that the appropriate suppliers had been paid.
- Financial stability of the battery system vendor doesn't end when the system is actually fielded. Typically, the first year of system operation will include some on-site repairs, spare parts, etc., and these are usually covered under a warranty agreement with the vendor. However, if they encounter financial hardships, then this can impose serious delays on operations when problems arise. During this project, the vendor did declare bankruptcy, and was eventually purchased by another company. There were some delays, and some issues in transferring support and warranty agreements, but fortunately all parties were committed to the success of the project and these issues were resolved fairly quickly.
- Proper selection and tuning of parameters are needed for optimization and design of the Phase Locked Loop (PLL).
- Controllers need to be optimized to increase the Short Circuit Ratio (SCR).
- The battery model used in the simulation needs to be validated either by experimental testing of a battery cell in the laboratory or by obtaining the information from the battery manufacturer.
- The controllers need to be optimized to allow the battery respond to the to the wind profile depending on the generation and load demand.
- The controllers should also be optimized to allow the battery respond to the under frequency events on a battery – wind connected grid system.

- The optimized controller for the battery management system should be able to respond to the frequency deviations thereby providing ramp support to stabilize the grid.

9. FAST RESPONSE REGULATION SERVICE WITH FLEET ELECTRIC VEHICLES

9.1 Introduction

Frequency regulation is a grid ancillary service that allows electricity generators, including those with large battery energy storage systems, to bid services into the energy market to assist in maintaining the electric grid frequency at 60 Hz. Providers of this type of service guarantee to be able to adjust generation during a specified time period by either ramping up or ramping down production according to instructions in order to maintain the balance of generation/load on the grid, and thus the frequency. Traditionally, in the ERCOT market, this was accomplished through the dissemination of a four-second signal, with each frequency regulation generation resource then responding within their pre-established ramp rate until they met the full capacity specified in the signal. Participants in the frequency regulation market get paid whether they are called upon to adjust power or not, thus they are paid to be “on standby”. As shown in the figure below, ERCOT studies (data is from 2008) have determined that the need for fast response resources increase as the concentration of wind grows on the grid. This project was developed to demonstrate the capability of addressing frequency regulation through the use of aggregated electric vehicle (EV) fleets to provide one type of fast response resources.

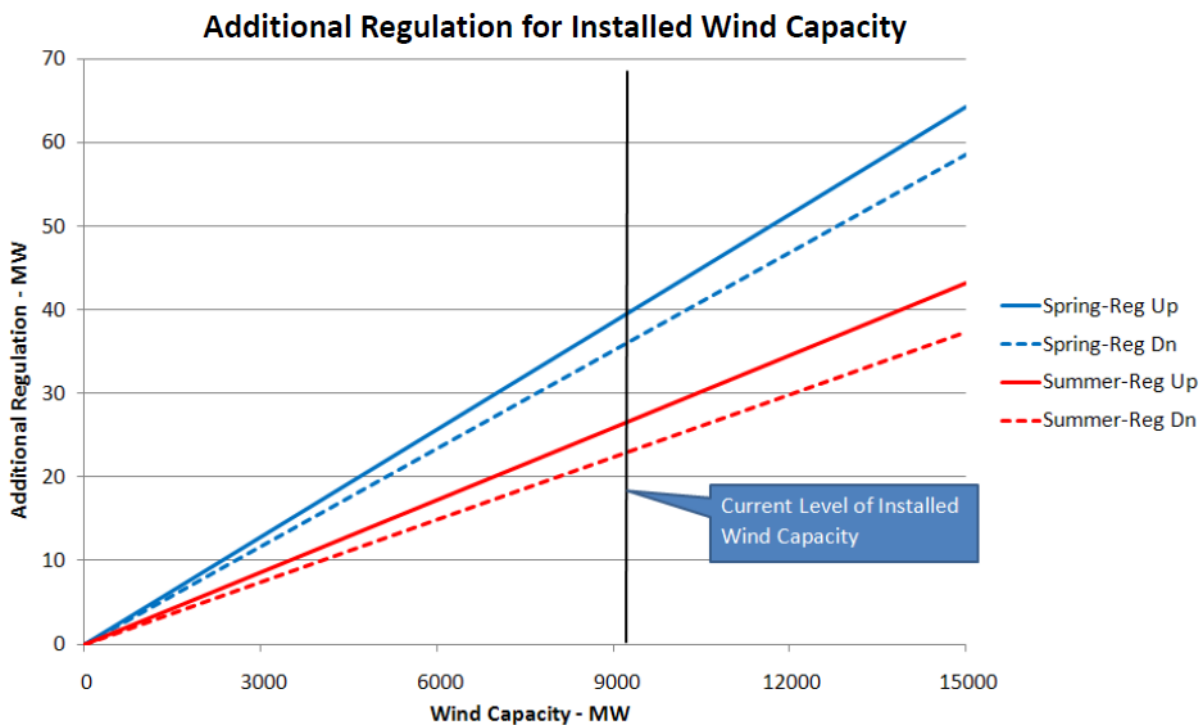


Figure 201. Results of a GE Study on Regulation Service Assuming 5-Minute Nodal Dispatch ⁴¹

In January 2013, ERCOT began a new pilot program to provide fast response regulation service (FRRS). In this type of frequency regulation, signals from ERCOT are responded to at full power within one second of signal receipt. Additionally, participants in FRRS monitor grid frequency, and can respond automatically when a deviation of .09 Hz or greater is detected. ERCOT's goal is to develop regulation services that can respond faster to grid frequency deviations, thus saving both in reduced congestion (and its associated costs to the market) and in reduced frequency regulation services since it won't require as many participants standing idle and ready to provide frequency regulation.

Battery energy storage systems can serve as an ideal provider for FRRS since batteries have a much faster response time than gas or coal-fired generators which must ramp to the regulation goal based on the simple physics of heating or cooling boilers. A great resource for batteries that could serve this purpose without significant investment (such as a large-scale utility scale battery costing millions of dollars) is through the use of EVs. In particular, supporting regulation-up services (the reducing of load on the system) can be as simple as "flipping a switch", i.e. turning the EV charger off. Regulation-down (the addition of load on the system) could be provided almost as quickly although the delivery of load to the grid would be managed by the EV battery management system (BMS) and so would be slightly slower.

⁴¹ Analysis of Wind Generation Impact on ERCOT Ancillary Services Requirements, http://www.uwig.org/attachb-ercot_a-s_study_final_report.pdf

The purpose of this part of the CCET project was to demonstrate an FRRS capability using an EV fleet at the Frito-Lay Distribution Facility in Fort Worth, Texas. Frito-Lay has a fleet of 11 EVs that are used primarily at night for deliveries around the Dallas/Fort Worth metroplex. These vehicles are typically charged during the day, when grid loads are highest and the need for frequency regulation is most apparent. This effort entailed the development and demonstration of an EV aggregation control system to support regulation-up services in response to either an ERCOT signal or a detected deviation of grid frequency of greater than .09 Hz. FRRS services were then bid into the market in one-hour intervals that matched the times the vehicles were drawing a load from the grid (charging), thus providing an excellent source for power reduction to support FRRS. The architecture for this effort is shown below.

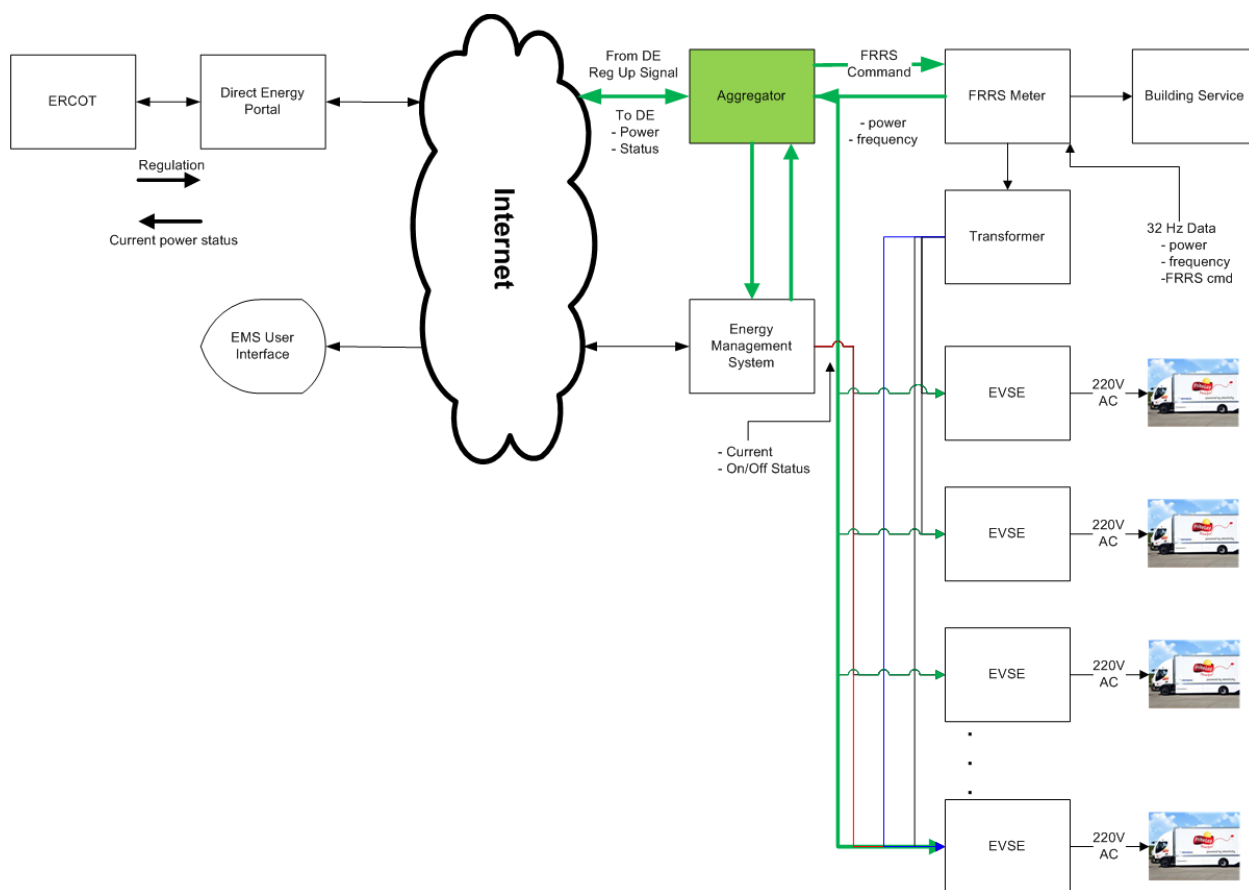


Figure 202. Architecture for Delivering FRRS from the Frito Lay Facility

The EV fleet at Frito-Lay successfully bid into the ERCOT FRRS market and received compensation for providing energy capacity to the grid. This became the first time EVs have successfully participated in an ancillary service in ERCOT and the first time a fleet owner has received payment for providing these types of services using an EV fleet.

9.2 Anticipated Benefits

Broadly, the expected benefits of the CCET demonstration project will be derived from a more reliable electric grid that can facilitate effective management of and responses to increased wind resources in Texas and from supporting the deployment of new products, technologies, and infrastructure to help customers make informed decisions about their energy usage. Specific smart grid benefits that are potentially supported by the FRRS component include:

- Enhanced grid capability to manage wind variation onto the grid by using temporary energy storage services as provided by EV fleets to help manage frequency deviations caused by wind.
- Pollution reduction as a result of decreased fossil-based energy production by enabling increased use of wind generation and then using EVs as a mechanism to manage frequency to meet short-term grid reliability needs.

9.3 Interactions with Key Stakeholders

The team worked closely with ERCOT and the qualified scheduling entity (QSE) in developing and implementing the bid strategy, and in compliance and validation of the actual regulation service. The Texas TDUs were informed of the project results, and other TDUs and ISOs have expressed an interest in learning more about the potential demonstrated by this project.

9.4 Technical Approach

This section provides a detailed description of the technical approach. The first part, Project Plan, identifies the steps taken to demonstrate the technology. The second part, Data Collection and Benefits Analysis, defines the technical approach for conducting the benefits analysis, including the methodology and algorithms when appropriate.

9.4.1 Project Plan

This effort consisted of design, development, integration, and deployment of technologies to utilize an existing EV fleet to deliver FRRS. The following activities were conducted to create and ensure a successful deployment of technology:

- Identification and procurement of a utility revenue-grade meter to monitor power and grid frequency at a minimum of 32 Hz. After an extensive trade study of suitable meters to meet the requirements of the electrical infrastructure at the site, it was decided to build a custom power meter. The electrical infrastructure would have required either a complete replacement of the electrical panel box to split off building power from charger power or the acquisition of 12 revenue-grade power meters. Designing and building a custom meter that could monitor the 12 EV circuits after the panel box became the more cost-effective solution. The solution used the National Instruments suite of components, including a CompactRIO DAQ computer, with

associated voltage, current, and relay modules. Total hardware costs were under \$20,000 versus over \$100,000 that would have been spent on individual power meters or redesigning the electrical panel box. The solution also allowed for a minimally-invasive installation suitable for future deployments. A further advantage was in data collection speed; this system exceeds IEC and ERCOT requirements for metering. Finally, with data collection and aggregation logic co-located on the same hardware, the processing and response times were increased since all data was flowing at hardware speed without going through communication interfaces. The final solution, shown below, resulted in a compact, robust, fast, and inexpensive power meter/aggregation system package that is easy to install and remove without requiring extensive site improvements to the electrical infrastructure.



- Development of a software aggregation control system to manage an EV fleet in accordance with FRRS parameters.
- Establishment of an interface to an existing on-site energy management system (EMS) that reported EV status information. After getting well into design of the system, it was decided that an interface with the on-site EMS was not necessary to accomplish the goals of the project, and so this interface was not developed.
- Establishment of a secure physical network interface to the ERCOT system in accordance with FRRS pilot requirements. In accordance with ERCOT rules, this interface was done through a QSE. A DNP3 interface to the QSE RTU was implemented over a secure VPN tunnel. The QSE telemetered the information provided by the aggregation system to ERCOT on a 2-second interval. ERCOT verified they were receiving the telemetered data from the aggregation system.
- Development of the data collection, correlation, and delivery to ERCOT of the required meter data files within one day of each frequency regulation bid period. The eventual solution to this was somewhat manual due to the rules for a QSE. After each bid hour, data files were automatically downloaded through a remote agent to SwRI. The data files were converted to a Microsoft Excel format, and then emailed to the QSE. After review, the QSE then sent the files to ERCOT. This solution worked well for the pilot and could be further automated in a future deployment if desired.

- Development and deployment of hardware for controlling the EV chargers within the FRRS established time frames for power responses.
- Installation and deployment of secure networking hardware for the interface between the QSE and ERCOT.
- Laboratory testing of all interfaces, using simulators, to verify communications pathways, data receipt and responses, and timing to enact FRRS power changes on the EV fleet, prior to deployment.

Success in this project revolved around a few key data items:

- Ability to monitor grid frequency and detect deviations of .09 Hz from the base 60 Hz. The final system detected frequency deviations of .001 Hz. The deployment grid frequency is configurable and can be changed while the system is running.
- Ability to implement FRRS protocol within 60 cycles (one second) of either a detected deviation or an ERCOT instruction. The final system repeatedly and consistently responded within 30 cycles (500ms).
- Ability to monitor and log fleet power consumption at a minimum of 32 Hz during an FRRS bid period. The final system monitored and logged power data at 40 Hz.

9.4.2 Data Collection and Benefits Analysis

Data collection focused on the requirements of ERCOT to provide verification of the ability to provide services that were bid into the ancillary services market. This included the monitoring and recording of grid frequency at greater than 31 times a second, amount of power being consumed by the EV fleet, amount of power being managed in accordance to the bid into the FRRS market, reaction times to grid frequency events, and EV fleet behavior during, going into, and coming out of a grid frequency event.

To satisfy the ERCOT data collection requirements, a data acquisition (DAQ) system was developed that monitored the electrical distribution circuit to each EV charging station. The DAQ collected voltage and current from each circuit every 25 ms, or approximately 40 times per second. The readings were summed to provide the total power consumption of the fleet. Additionally the current readings were combined with one voltage reading to compute grid frequency using a Fast Fourier Transform algorithm. The EV aggregation system also computed the bid capacity by summing the power readings from the subset of chargers that were computed to be participating in the bid since the last 25 ms increment. The combined data readings, as well as the computed frequency were logged into a telemetry file every 25 ms. These files were saved and downloaded from the system after each bid hour.

Methodologies to determine the direct benefits of this system to increase wind energy reaching the grid are beyond the scope of this project. The primary focus was to prove that EVs can be used with the grid in a manner that would increase wind energy use. Those benefits are realized by a fleet that can successfully regulate their charge rate in accordance to grid frequency, thus preventing the need to provide additional power plant generated energy to provide additional energy into the grid. While the

amount of wind energy actually on the grid doesn't change during a frequency decline, the amount of traditional power plant generation is affected in that it is not increased when frequency declines. Instead the amount of load represented by the vehicles is removed from the grid, freeing up that energy for other uses and maintaining the percentage of wind versus traditional generation that is currently on the grid. This means that more energy consumed by loads on the grid are satisfied by wind, and that less controllable energy generation from gas and coal plants is needed to supply that load. This in fact increases the consumption of wind compared to traditional, and reduces emissions since the traditional plants don't need to increase production and thus emissions.

9.5 Performance Results

This section covers the operation of smart grid technologies and systems, including the demonstration results; impact metrics and benefits analyses; and stakeholder feedback.

9.5.1 Operation of Smart Grid Technologies and Systems

The final system for FRRS fleet management was an autonomous system that ran 24/7/365. Each week, the bid schedule for the following week was uploaded into the system, setting its bid calendar. When a bid hour began, the system started monitoring power and computing frequency, deploying vehicles as needed to satisfy the bid. When the bid hour ended, the system entered an idle state until the next bid hour began.

The system consisted of three operational components:

1. Power monitoring and control system
2. Aggregation system
3. Remote access and data management system

For the purposes of this system, the following terms are defined:

- Deploy or deployment: turning off the EV chargers so that power is not able to flow to them.
- Undeploy or Undeployment: turning on the EV chargers so that power is able to flow to them.
- Setpoint: the lower and upper frequencies as determined by ERCOT for deploying and undeploying assets. For this pilot the deployment frequency was 59.98 Hz, and the undeployment frequency was 59.99 Hz.
- Recall: same as undeploy.

The power monitoring component used a field programmable gate array to collect power values from each EV charger circuit every 25 ms, provide summated fleet power values, and compute grid frequency. These values were fed to the aggregation system. The control system component was listening for aggregator commands to deploy or undeploy vehicles. Upon receiving an aggregator command, the control system sent commands to the appropriate relays that then sent an electronic signal to the

charger contactors to either open or close them, in effect “flipping the switch” on them to turn them on or off. The power monitoring system was also responsible for recording the data during bid hours for ERCOT validation.

The aggregation system contained the logic for managing the EV charger fleet in accordance to the bid schedule and the grid frequency. The aggregator’s first task was to determine if the system was in a bid hour. If so, then it next checked to see if it was already in a deployment to determine whether it should compare frequency to the deployment setpoint or the recall setpoint. It then looked at frequency and determined if it should issue a command to the chargers. Finally, if it was undeployed, in a bid hour, and not at a deployment setpoint, the aggregator created a prediction of which chargers should be turned off in the next 25 ms increment of time, based on their individual power readings and the amount of power bid into the market. The aggregator ran on a 25ms loop time to align with the power monitoring and control system. Its prediction algorithm provided the decision speed needed for deploying within 60 cycles of a frequency deviation beyond the setpoint. To summarize, the sequence of decisions made by the aggregator were:

- Determine if the current system time was within a bid hour
- Determine if the system was currently in a deploy mode
- Determine if frequency required a command and issue command
- Predict the next deployment if not currently in a deployment

If the current system time was not in a bid hour, then the aggregator sat idle.

The third component, remote access and data management, was a web-based component that allowed for remote management of the system. The entering and uploading of bid schedules was done through this component. Downloading of data files occurred automatically after each bid hour through the remote system. Changes to the configuration of the system were managed through the remote system, and could be uploaded and implemented while the system was running. Synchronization between the remote data directory and the local data directory was also managed by the remote system to ensure that data files were successfully downloaded before being deleted from the deployed system. With this component there was no need to access the deployed system locally; all system and data management was handled by the remote client. The remote access system used an SSL connection over a private VPN to secure the data transactions and prevent unauthorized access to the aggregation system. Access to various reporting and troubleshooting logs was also done through the remote access system.

9.5.2 Metrics and Benefits Analysis

After the project was completed, some specific metrics and benefits were defined. This section provides some insight into what those metrics and benefits are. They primarily revolve around two key concepts: 1) ability to participate in FRRS according to the rules outlined by ERCOT, and 2) the economic benefits realized from using an EV fleet for FRRS. The operational metrics include:

- Power response to FRRS signals. This metric measures the EV fleet response to an FRRS signal based on an appropriate power level and maintaining that power level for the defined time period without significant deviation.
- Frequency deviation detection. This metric involves capturing frequency from the grid at 32 Hz sampling rates and the detection of deviations at a resolution of 0.01 Hz.
- Time to implement FRRS signal. Participation in the ERCOT FRRS pilot requires implementation at either the specified power level or at full offered power with 60 cycles (1 second) of either notification via ERCOT signal or detection of frequency deviation through power line monitoring. This metric captures the implementation time.
- Percent of successful deployments in offering period. This metric indicates the success rate for meeting deployment obligations during the offering period (1 hour). The metric combines the ability to implement within 60 cycles and the ability to sustain deployment to within 95% - 110% of obligated capacity, both of which are required for successful response.
- Percent of successful deployments during pilot. This metric compares the results of responding to all of the offering periods during the pilot.
- Economic impact or benefit to fleet owner. This metric compares the revenue generated through participation in FRRS versus any penalties that would have been assessed by ERCOT as a result of unsuccessful deployments. This metric was compiled for each offering period, and for the pilot period.

9.5.2.1 Operational Metrics and Benefits

Ultimately, the economic benefit to the fleet owner will be defined based on not only potential revenue but also considering the costs to configure a fleet for participation and any operational impacts as a result of participation in the FRRS offerings.

These next sections discuss the defined metrics. Data was collected from over 150 simulated and actual bid hours, representing over 21,500,000 data points taken between December 11, 2013, and February 28, 2014, the end of the ERCOT pilot. Simulated bid hours represent where the system was set to deploy 10 kW of power, allowing the data collection, computation, and recording components of the system to run without requiring a large number of vehicles to be on the chargers. Actual bid hour represents where the entire fleet was available and the deployment power was set to 100 kW, the minimum that could be bid into the ERCOT market. As it turns out, 147 of the 150 hours were simulated. The issues causing this are discussed in the lessons learned section of this document.

9.5.2.2 Power Response to FRRS Signals

For each simulated and actual bid hour enacted by the system, the system responded to 100% of the frequency deviations at the specified deployment setpoint. Each frequency event, defined as when the frequency declined below 50.98 Hz, was responded to by the system recognizing the frequency and sending deploy signals to the chargers. The system recalled from a deployment 100% of the time when the frequency subsequently increased to above 59.991 Hz. In no instances was it found that the system

did not maintain a deployment for either the entire frequency decline event period, or the end of the bid hour, whichever came first. For deployments that crossed hour boundaries in back-to-back bid hours, the system successfully maintained the deployment across the hour mark without deviation.

9.5.2.3 Frequency Deviation Detection

Each 1-hour data file that was collected by the system was found to contain 144,000 data readings. This correlates to 1 reading every 25 ms, or a data collection rate of 40 Hz, exceeding the requirement for 32 Hz data capture. The system records frequency at .001 Hz precision, exceeding the standard of .01 Hz. This was calibrated in the lab using a calibrated signal generator. During system qualification time and actual bid hours, system data files were validated by ERCOT using their readings and were found to be within expectations with their frequency readings. It should be noted that frequency is not consistent at the .01Hz level across the grid and that readings taken in any one location can be affected by local variations in the grid, such as nearby heavy equipment like arc welders or dynamometers. In fact one aspect of the FRRS program was to automate the deployments based on local frequency readings in order to increase the speed in which resources react and more quickly mitigate the effects of frequency variation than can be done through detection and dispatch from a central point.

9.5.2.4 Time to Implement FRRS Signal

The requirement was to implement an FRRS deployment within 60 cycles of detecting a frequency decline where frequency dropped below 59.98 Hz. The system successfully deployed to each frequency event within 450-600 ms 100% of the time.

9.5.2.5 Percent of Successful Deployments in Offering Period

For each deployment power specified (either 10kW or 100kW) the system was able to achieve 100% deployment of whatever capacity was available within 60 cycles with a deployment power level of 95% - 110% percent of obligated capacity. It should be noted that most of the hours were simulated at low deployment power levels (typically at 10%, although other percentages were also used in simulation). The system consistently deployed either maximum available capacity or obligated capacity in the 95% – 110% range, whichever was lower.

As it turned out during this pilot period, there were extensive mechanical problems with the vehicles that could not be fixed due to the vehicle vendor not honoring maintenance commitments in a timely manner. This caused issues that prevented full obligated capacity being available, requiring the actual bid hour to be converted into a simulated hour.

9.5.2.6 Percent of Successful Deployments During Pilot

Because of ongoing maintenance problems with the EV fleet, only 3 hours within the pilot became actual bid hours, and of those 3 only 1 hour was completely successful (shown below). For the other

two hours, it was determined in a post-mortem analysis that fleet processes changed the parameters of the fleet. For example, on one day, an ice storm hit the Dallas/Ft. Worth area and so several of the trucks did not go out for their runs. As a result they were fully charged when the bid hour arrived and thus were not drawing charge from the grid, resulting in total power consumption for the fleet less than 100 kW, the minimum obligated capacity for FRRS services. While the system successfully deployed the remaining capacity (approximately 70 kW) during that bid hour, the fleet as a whole did not successfully meet the requirements for the bid hour since it did not show a deployment of 100 kW of load being removed from the grid during a frequency event. In the other 1 hour instance, a vehicle was charged in the shop during the day after its deliver run in order to be used for driver education. As such the vehicle achieved full charge within 10 minutes of the bid hour and then stop charging, reducing the total fleet capacity to below the 100 kW threshold. Since 2 of the other 12 vehicles were out for maintenance during the entire pilot time, having even 1 vehicle that did not adhere to its normal schedule took the fleet out of consideration for deployment.

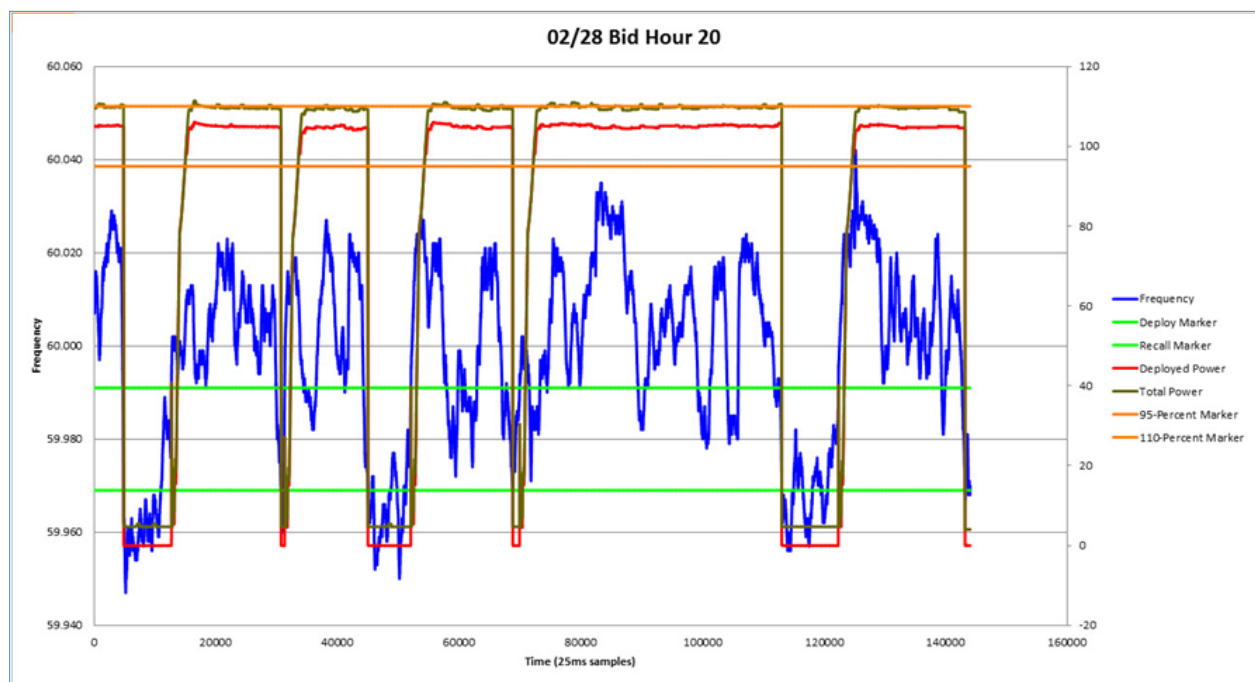


Figure 203. Successful Deployment of FRRS

9.5.2.7 Economic Impact or Benefit to Fleet Owner

In the ERCOT market, given the current market rules, and with energy being relatively inexpensive in Texas, there is little to no economic benefit that can be realized by small fleet owners by participating in the FRRS market. As a part of this project, the Team looked at ancillary service prices paid for frequency regulation services in ERCOT for 2013 and up through March 2014. Average frequency regulation prices paid per 100 kWh, by hour, for 2013 are shown below.

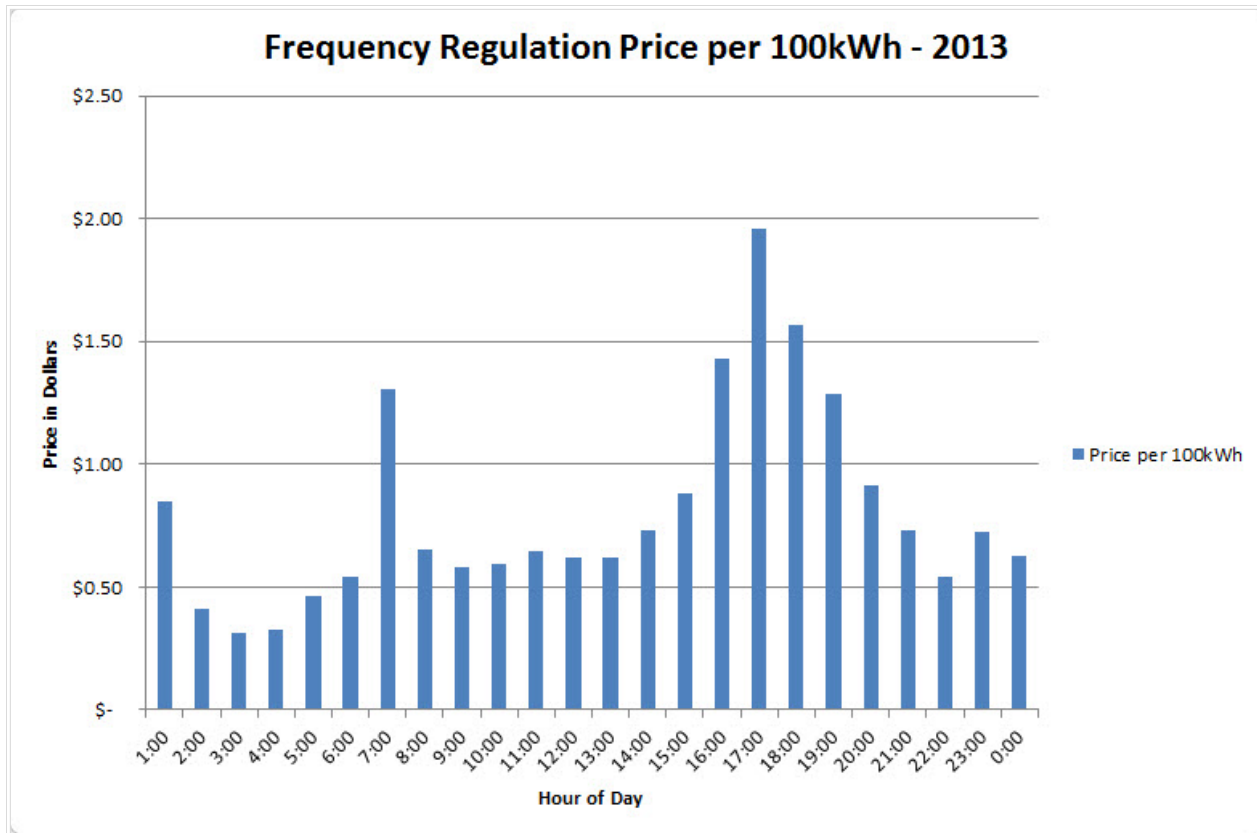


Figure 204. Average Frequency Regulation Price Per Hour Per 100 kWh in ERCOT for 2013

In 2013, there was a specific peak in frequency regulation pricing in the 6:00 a.m. - 7:00 a.m. time period, and a broader peak from 3:00 p.m. - 8:00 p.m. For the fleet in this project, the broader peak aligns well with the typical idle period of the vehicles, which do their deliveries between midnight and 10:00 a.m. and are loaded during the afternoon hours. Starting at around 5:00 p.m. they are all sitting on the chargers loaded and waiting for their next delivery run. The average price per 100 kWh, however, is relatively low, around \$.80 - \$1.95 per 100 kWh. Looking at a potential of 4-hours per day that the project fleet could realistically and reliably participate in the FRRS market (from 5:00 p.m. – 9:00 p.m.), the following table shows that for 2013 the average potential monthly revenue for a 100 kWh fleet was \$135.01.

Table 55. 2013 Potential Revenue from Participating in the 5:00 – 9:00 p.m. Timeframe

Hour Ending	\$ per 100 kWh	30 Days \$
18:00	\$ 1.57	\$ 47.01
19:00	\$ 1.28	\$ 38.50
20:00	\$ 0.92	\$ 27.50
21:00	\$ 0.73	\$ 22.00
Total		\$ 135.01

Up through March 15, 2014, as shown below, the peak moved substantially, and also the price for frequency regulation services increased significantly.

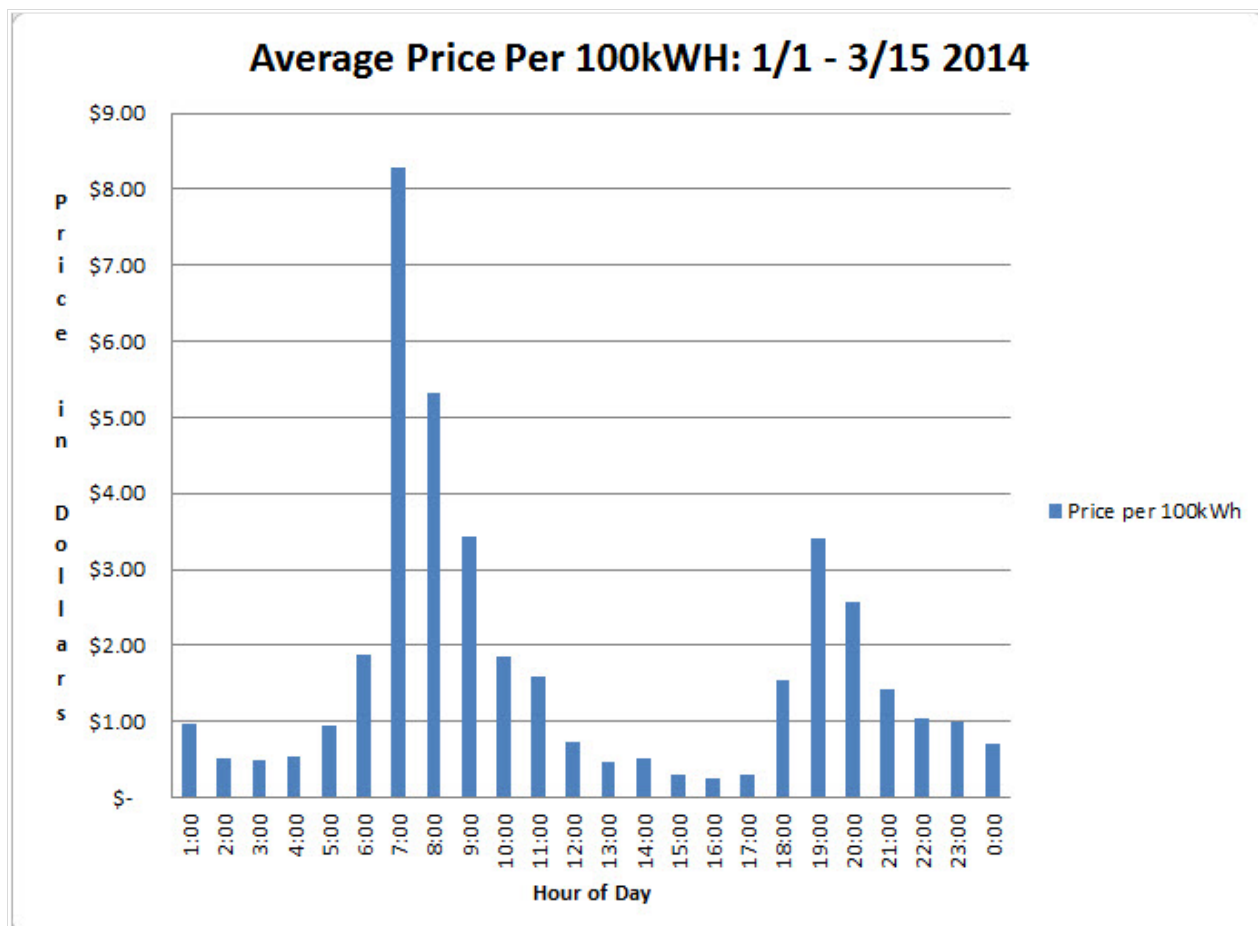


Figure 205. Average Frequency Regulation Price per Hour per 100 kWh in ERCOT for 2013

While the morning opportunity was not available for the project fleet due to their delivery schedules, the increase in price does have a positive effect on potential revenue, as shown below.

Table 56. 2014 Potential Revenue from Participating in the 5:00 p.m. – 9:00 p.m. Timeframe

Hour Ending	Price per 100 kWh	30 Days \$
18:00	\$ 1.55	\$ 46.53
19:00	\$ 3.40	\$ 101.97
20:00	\$ 2.58	\$ 77.27
21:00	\$ 1.42	\$ 42.51
Total		\$ 268.29

The data shows that for 2014 the potential revenue was double that which was achievable in 2013. Even so, this was not sufficient to offset the costs of equipment, installation, and ongoing expenses. For example, all assets must go through a QSE to participate in market services, so there is a monthly price assessed by the QSE for maintaining the communications links to the asset and for representing them in the market (placing bids). The price for the QSE services for this project was \$2,000 per month.

In order to be profitable in the current market structure, the fleet would need to be of an approximate capacity size of 1 MWh, however that is very unlikely. Alternatively, if the fleet could be flexible in its scheduling, it could place bids in more opportune time periods, reducing the size of the fleet needed to participate successfully from 1 MWh to 400 kWh, as shown below.

Table 57. 2014 Potential Revenue from Participating in the 6:00 a.m. – 11:00 a.m. Timeframe

Hour Ending	Price per 100 kWh	30 Days \$
6:00	\$ 1.89	\$ 56.63
7:00	\$ 8.28	\$ 248.53
8:00	\$ 5.33	\$ 159.89
9:00	\$ 3.42	\$ 102.72
10:00	\$ 1.86	\$ 55.75
11:00	\$ 1.60	\$ 47.91
	Total	\$ 671.43

There are also the potential jackpot scenarios, catching an hour where the price of energy skyrockets due to grid congestion. While this effort was operational in 2014 there were at least 4 such hours where the cost per MWh ranged from \$284 to \$4,999. Those hours were normalized in the calculations to remove skewing of the data.

This project, as stated earlier, was conducted in a market where energy is relatively inexpensive and plentiful, and the market is structured for large, stationary generation resources. In other markets, such as California where the price of electricity is significantly higher, the economics would be much more favorable.

9.5.2.8 Build Metrics

Build metrics are related to equipment types, numbers and costs. As identified earlier, the Team purchased various NI hardware and software to implement the FRRS capabilities, and these purchases amounted to about \$15,579.

9.5.2.9 Impact Metrics

Impact metrics are related to reductions in energy use which also drive reductions in pollution emissions. One could presume that using fleet EVs to reduce the need for traditional generators

providing regulation service would also reduce fossil-based fuels and hence pollution emissions. However, the Team has no information relative to the types and sizes of potential reductions in traditional generation that could be achieved by using EV fleets to provide FRRS.

9.5.3 Stakeholder Feedback

ERCOT validated the successful delivery of FRRS using the fleet electric vehicles, and referenced the participation of this project in a review of its FRRS pilot ⁴².

9.6 Conclusions

This section covers projections of demonstration and commercial scale system performance as well as lessons learned and best practices.

9.6.1 Projections of Demonstration and Commercial Scale System Performance

From a technology perspective, the technology built to enable the fleet to participate in FRRS services according to the requirements of the FRRS program works reliably and consistently. From installation in early December 2013 through the end of February 2014, the EV aggregation system ran 24/7 without missing a reading or event. The aggregation system was built with future deployments in mind and is small, self-contained, requires minimum disruption of the existing electrical system, and is built of low-cost but highly reliable and accurate off-the-shelf components. Since the system was built with the same components used for other smart grid applications such as synchrophasor units it is expected that it would continue to run for many years without interruption in normal operating conditions. The system is easily replicable; new systems for managing up to 12 vehicles could be created and installed in 1-2 weeks, plus shipping times for ordered components. Enabling the system for larger vehicle fleets would require additional hardware and some minor coding changes, but could be easily and inexpensively done. What would be beneficial is to estimate how large of a fleet would be required for the effort to be profitable. Since the aggregation costs are not very sensitive to scale, and the QSE costs would probably be the same regardless of scale, such an addition would be informative.

9.6.2 Lessons Learned and Best Practices

While executing this project, the Team did learn some lessons and best practices, and those are shared herein.

⁴²

http://www.ercot.com/content/meetings/tac/keydocs/2014/0327/11._TAC_Final_Report_on_FRRS_Pilot_as_of_3-25-14.ppt

9.6.2.1 EV Fleet Size for Adequate Market Participation

This pilot was conducted using a vehicle fleet of 12 vehicles and 12 chargers. Each vehicle consumed between 9-11 kW during normal charging, for approximately 4 hours, tailing down in the last hour as the battery reached full state of charge. The minimum power requirement for participating in the ERCOT FRRS pilot was 100 kW, meaning that the program had two additional vehicles over the minimum limit in which to satisfy a bid hour. The project could have had 3 good bid hours each day based on the above conditions, but the Team learned that having 100 kW of power available was a challenge due to several unforeseen factors.

One of the primary reasons for not always meeting the 100 kW power limit was vehicle maintenance issues. These vehicles are an older generation model, so there were issues with the vehicles requiring maintenance, and not just typical hardware mechanical issues; several vehicles had software issues that affected their ability to accurately charge. Therefore, at times, there were only 7 - 8 vehicles available so the fleet could not deliver sufficient power to satisfy the FRRS delivery requirements.

9.6.2.2 EV Fleet Behavior and Processes

Another issue to be considered for the future is fleet behavior and processes. This project had access to extensive vehicle route data and vehicle charge data, and was able to develop a bidding scheme for when the vehicles would be optimally on the chargers pulling full charge. Processes at the facility, however, frequently interrupted the planned schedule. Issues such as shop charging (using a non-aggregated charger in the shop to charge a vehicle) for driver education or after scheduled maintenance would result in a fully charged vehicle being in place during the bid hour when the expectation was that it would need to be charged. Weather, which prevented some vehicles from making deliveries, caused problems because the vehicles were not discharged as anticipated. Loading schedules were occasionally flexible, resulting in vehicles being removed from chargers when they were expected to be available. There were also vehicles left at the loading dock for extended periods after they were loaded instead of being returned to the chargers. All of these typical operational incidents demonstrated that analyzing just fleet route and charging statistics is not sufficient for planning participation in an ancillary service market, but rather it is important to understand the fleet processes so that behavior can be incorporated into the equation.

9.6.2.3 Market Rules for Participation in Ancillary Services

The current ERCOT market rules are designed for stationary, large scale generation facilities where generation is their primary, and in many cases, only mission. ERCOT did define a lower power capacity limit for participation in the FRRS service (100 kW vs. 1 MW normal) pilot, however having to meet a defined minimum power level with aggregated small scale assets can be a real challenge.

Compounding this challenge is the requirement that all bids be entered into the market 1 week in advance, meaning that a fleet owner has to predict the status and availability of their vehicles for the

following week. Considering the nature of the vehicles, and all the things that can happen to a vehicle when it is driven (from maintenance to accidents to not being used on a given day), it would be best if a fleet owner has a significant over-capacity on-hand (in the order of 50%) in order to meet unforeseen events. Even then that does not guarantee sufficient availability; it just statistically minimizes the more common problems that can occur.

The market structure for frequency regulation is first and foremost a pay for capacity, where market participants are paid for the capacity they make available. What they are really paying for is idle time of the asset, assuming that it is doing nothing during that period of the bid and so the asset owner is getting paid for making the asset available “in case it is needed”. FERC Order 755 attempted to introduce an additional payment, pay for performance, in order to further reimburse assets that are actually used in an event. However, pay for capacity remains the primary reason for making an asset available for frequency regulation. For the EV fleet owner, this means an adequate number of vehicles must remain in position and be available just in case their service is needed during that bid timeframe.

Two recommendations are put forward for future EV fleet implementations for FRRS or other types of capacity-based services.

1. Consider over-sizing the participating fleet to incorporate the normal (90%) cases that can reduce fleet size for the bid duration due to maintenance, accidents, use outside of normal practices, or other statistically predictable conditions.
2. Change the market rules to pay EV fleets (or other aggregated small scale assets) based on performance without the requirement for a known capacity size to be specified in advance, and allow for shorter duration delivery of services.

Over-sizing, while computable based on typical statistics methods and databases on vehicle performance and behavior, would limit fleets from participating unless they were of substantial size. For example, in the ERCOT market which has a limit of 100 kW, an oversized fleet where vehicles charged at a normal 10 kW would be around 15 vehicles, in order to ensure that 10 would be available for any bid hour to meet the capacity requirement.

Changing the market rules would require two fundamental changes: changing the pay structure to allow for increased pay for performance as the only payment to EV fleets or pay for capacity for dynamically-sized power assets like a fleet; and increased telemetry to the ISO/RTO to know what is available in any given moment for the service.

A change in the payment structure could allow for increased participation of EV fleets in the market. Currently the payment structure in ERCOT is for an hour; however, it could be reduced to 5 minute increments based on telemetered data. This would allow the fleet to be able to participate for more hours of the day, and participate every day no matter what their vehicle usage is.

Another way the payment structure could be changed would be to pay for performance only, increasing that payment to justify participation. In this instance the fleet owner only gets paid for assets that actually get deployed, but at a rate that incentivizes making them available whenever possible. Telemetering the current fleet size data to the ISO/RTO would allow them to know what is available at any given time for their short-term planning purposes.

The primary argument for a dynamically-sized asset would be the ISO/RTO's need to know what assets they have available to respond to an event, particularly an N-1 event. Because of the number of assets there are in a grid, the computations to arrive at the next dispatch decision take several minutes and so cannot or do not run in true real time. This project has shown that the fleet capacity does not change much during the bid timeframe (hour); rather it changes during the day or days preceding such that it can be significantly different from the bid prediction made a week earlier. Minor changes, such as a vehicle plugging-in or being unplugged, can be statistically calculated and managed in the wider picture of the whole grid.

9.6.2.4 Market Rules for Bidding into the Market

The current process that was used for the ERCOT FRRS pilot was similar for other frequency regulation services designed for stationary assets; submit bids one week in advance for the following week. With a gas turbine generation plant that never moves and has a sole purpose of producing power, or a large battery system that is stationary, this should be an easy exercise. For a mobile asset that has a second primary purpose where lots of things can happen to the asset outside of energy generation, this becomes a difficult challenge.

The ERCOT FRRS pilot did allow for canceling a bid 24 hours in advance, but even that is not always sufficient. For example, the fleet owner bid hours were in the evening. During the day, one of the vehicles was charged and used for driver training after its route. The training did not drain the battery nearly as much as the normal delivery route does and as a result the vehicle reached full state of charge during the bid hour and thus became unusable for the FRRS service. Here again, having extra capacity available to bid would resolve this type of situation.

Since EVs are dynamic energy assets, they need a more flexible process for bidding into the market. Some markets allow for bids 1 hour in advance. This would be a more appropriate method for EV fleet participation in FRRS with a better guarantee of success for the bid hour.

9.6.2.5 Market Rules for Connecting to the Market

In ERCOT, QSEs are the only ones who can actually bid into the market. As a result, with any intermediary in a process, there is a fee associated with their service. QSEs actually communicate with ERCOT through telemetry, and they are responsible for setting up, maintaining, and getting telemetry from any assets they manage. This means there is a set-up cost for a fleet owner to connect to the QSE, and ongoing maintenance/connection costs for the telemetry link. These costs can be excessive for

small fleet owners, and make it difficult to realize positive net revenue from FRRS. Until the conditions change, this will remain a significant barrier to participation by small assets like an EV fleet, especially in Texas where electric prices are relatively low.