

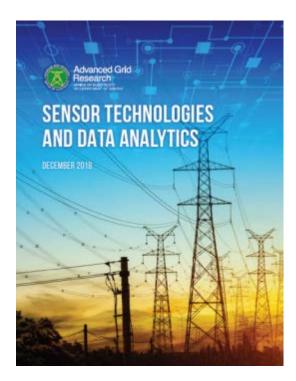
Data Analytics R&D within the DOE OE



Program Manager: Eric Lightner

March 2019

Sensor Technologies and Data Analytics Program



Core Technical Areas

- New Program for 2020 Enhanced Power System Resilience
- Incipient Failure/Fault Detection
- Detecting and Forecasting **Behind-the-Meter DER Impacts**
- Monitoring for Critical Infrastructure Interdependencies

Crosscut Technical Areas

- Collaborative, Cyberaware Sensors
- Sensor Valuation

Program launch anticipated in FY20 - contingent on appropriations

OE Priority Area and a Key Technical Area for GMI/GMLC Roadmap Implementation





Grid Modernization Laboratory Consortium (GMLC)



- A strategic partnership between DOE and the National Labs
- Brings together leading experts, technologies and resources
- Lab call funded 88 projects

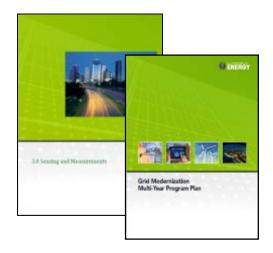
GMLC Projects laid the groundwork for new Sensor and Data Analytics Program





OE & GLMC Funded Projects Relevant to Analytics

- Discovery Through Situational Awareness (PNNL)
- Advanced Machine Learning for
 Synchrophasor Technology (LANL)
- Integrated Multi Scale Data Analytics and Machine Learning for the Grid (LLNL)
- Sensing Electrical Networks Securely and Economically (GA Tech)
- ADMS Test Bed VVO Use Case (NREL)





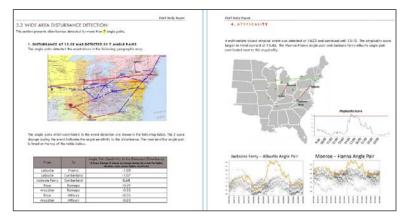


Discovery Through Situational Awareness

Project Goal: Creating situational awareness methods and tools that rely on machine-learning, statistical data mining, and data visualizations to provide insight into power grid behavior

ESAMS

Eastern Interconnection Situational Awareness Monitoring System



- Incorporates multivariate, ML algorithms for detecting phase angle anomalies and oscillations.
- Developing event detection and classification algorithms for load switching, faults, generator drops, & reactive load changes

MARTINI

Testing environment using real-time PMU data



• Evaluates anomaly detection algorithms for accuracy and functionality prior to being incorporated into ESAMS





Advanced Machine Learning for Synchrophasor Technology

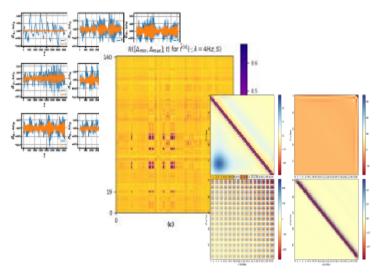
Project Goal: Use PMU data to develop Dynamic Machine Learning Technology for

- Parameter identification & topology change detection (under limited observability)
- Hidden anomaly detection such as asset malfunction (generators, transformers)
- Event/anomaly classification (load drop, forced oscillations, etc.)

Modern Data Science Tools:

- Graphical Model Learning
- Constrained Optimization
- Applied Statistics
- PCA, filtering
- auto- & cross- correlations
- Neural Networks (e.g. deep)







Physics (Power-Systems) **Informed Tuning**

(Power System interpretable but repetitive & off-line, hand-controlled)

PMU data-topredictions approaches



Physics Informed

Machine Learning

Physics-Free Machine Learning

(automatic, training & execution efficient, but lacking Power System interpretability)

Speed













Physics (of Power Grid) Informed Machine Learning

Estimator

 targeted prediction with enough of accuracy
 (in the focus of our efforts so far)

Reinforcer

• active exploration, e.g. for control, optimization (next stage)

Explorer

• model reduction on ML-steroids (working on it)

Accuracy



Speed









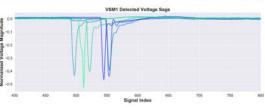
Integrated Multi-scale Data Analytics and Machine Learning for the Grid

Project Objectives: Develop advanced, distributed data analytics solutions; Provide visibility, and controllability to distribution grid and building operators, leveraging multi-scale data sets, from both sides of the meter; Evaluate and demonstrate machine learning techniques to create actionable information for grid and building operators.

Approach

- Investigate how machine learning techniques can be applied to streaming distributed datasets, & evaluate impacts of data quality on those methods
- Investigate how ML and data analytics techniques from other industries could be applied to the grid space
- Create an event signature database (Event Detect) to evaluate new/existing anomaly detection/classification algorithms and develop new anomaly algorithms for specific use cases
 Applications investigated
- ✓ Topology and parameter estimation
- ✓ DR and DER availability and verification
- \checkmark Event detection, classification and localization using streaming μPMU data









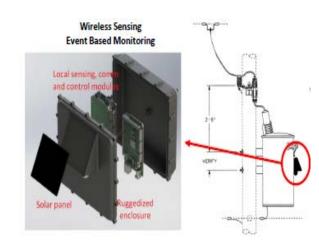
Georgia Tech Research Corporation

SENSE (Sensing Electrical Networks Securely and Economically)

Project Goal: Develop and demonstrate a secure, ubiquitous, low-cost sensor network for monitoring distribution assets which provides a sustainable ROI for utility deployments.

Project Features

- Incorporate modular sensors based on 3D printing for voltage, current and temperature, configurable for transformers, capacitor banks, reclosers and fuses.
- Integrate flexible energy management, communications, adv. functionality, data storage, cloud connectivity, cybersecurity, analytics, machine learning and user interface.
- Develop ML algorithms that identify symptoms of degradation based on measured current, ambient temperature and case temperature, for better prediction of transformer life
- Leverage historical Pecan Street AMI data to "learn" loading vs temperature relationship





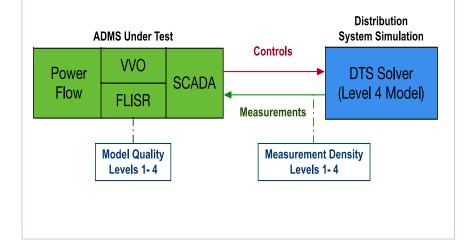


ADMS Testbed: Model Improvement Use Case

Project Objective: Evaluate the performance of an ADMS VVO application for different levels of data remediation and measurement density

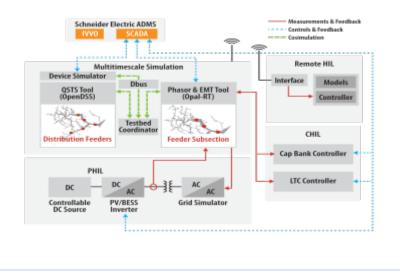
First phase:

- Software simulation only
- 4 levels of model quality
- 4 levels of measurement density



Second phase:

- Simulation on ADMS testbed
- Subset of scenarios







ADMS Testbed: Scenarios

Model Quality (MQ) Levels

- 1. Base level GIS data
- 2. Field verification at select locations; confirm step transformer attributes, and collect capacitor, regulator and recloser attributes
- 3. Tap phase verifications
- 4. Field confirming each primary pole line by circuit to obtain distribution transformer attributes, phasing, and using Xcel Energy GIS data

Measurement Density (MD) Levels

- 1. Feeder head measurements.
- 2. D1 + voltage regulators, capacitor banks, reclosers, and 1 tail-end voltage sensor (AMI sensor) per feeder
- 3. D2 + a total of 10 AMI sensors per feeder
- 4. D2 + a total of 20 AMI sensors per feeder.



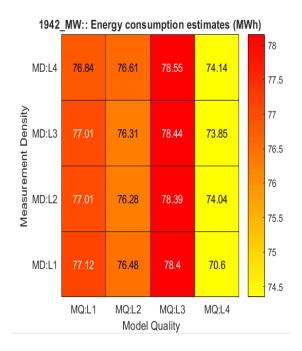


ADMS Testbed: Results

Summary of Findings

- Results vary across feeders
- Long feeders show difference in performance for varying measurement densities
- Rural feeders show greater improvement with increasing levels of model quality
- Higher measurement density results in fewer violations

Results for a semi-rural feeder







New Funding Opportunities

FY19 FOA: Big Data Analysis of Synchrophasor Data

- Explore the use of big data, AI, and ML technology and tools on PMU data for better grid operation and management to enhance the reliability and resiliency of the nation's power grid
- Pending selection notifications and award

FY19 GMI Lab Call: Advanced Sensors & Data Analytics

- One of five major areas
- Reflective of the foundational R&D topics identified in the Sensor Technologies and Data Analytics MYPP
- Addresses key OE priority
- Opportunity to jump start program prior to FY2020





Voices of Experience Series

ENERGY () ENERGY VOICES VOICES of Experience Integrating Intermittent Resources. **Voices Series** of Experience What Utilities are Learning. August 2012 ENERGY TRANSPORT Focus on areas that pose operational VOICES challenges or are of Experience VOICES critical to industry of Experience transformation. Insights into Advanced Distribution Management Systems Share knowledge to February 2016 further enhance grid A lines in A state in all \$ modernization 1 0 efforts.

Capturing the Collective Voice and Experience of Utilities at the Forefront on Modernization Efforts





Voices of Experience

- Analytics is the foundation for making broader data-driven decisions
- Finding hidden failures or patterns requires advanced analytics and pairing AMI with new data sources (e.g., data about weather, lightning strikes, etc.)
- Involves the analysis of large imblanced and multi-structured data sets
- Requires considerable computing power that can be costly to maintain
- Cloud computing raises privacy and security concerns and questions around cost recovery

AMI and Analytics

Using AMI for Analytics

- Theft detection/revenue protection
- Asset health and maintenance
- Predicting equipment failures
- Assessing impact of TOU rates
- Addressing power quality issues
- Understanding impacts of DER
- Increasing safety (detecting downed conductors & identifying unregistered DER)
- Improving customer service (new program recommendations or anticipating customer calls)





Voices of Experience

What Utilities are Doing

- Florida Power & Light Company: iOMS, tool-based artificial intelligence ticket processing robot that eliminates non-value added truck rolls. 96% accurate, runs 24/7, resolves tickets 8X faster.
- **KCPL/Westar Energy:** Piloted a predictive failure effort for transformers. Allows them to reduce unanticipated transformer failures, better plan replacements, reduce outage times, and decrease overtime costs.
- **Oncor:** Outage notification platform sends proactive messages to customers when an outage is detected, includes an estimated time of restoration and notifies customers when power is restored.
- **PG&E:** Used customer analytics to target communications to customers that would be the most receptive to a product or service; increased enrollment in their Solar Choice program from 1 to 3%.

Insights

- Data is a valuable asset
- It's not necessary to start with a big program – start small and expand
- Pair someone who understands the data with someone who knows the business
- Requires different staffing resources
- Data scientists can use analytics to find information hidden in the data
- Look outside the utility industry at nontraditional disciplines
- **SMUD:** Will use segmentation and customer analytics to better understand forecasted load reductions based on demographics and house size for improved distribution planning forecasts





Overview



DOE facilitated an industry-developed framework for the protection, access, use, and sharing of customers' electricity usage data.

- A voluntary code of conduct for energy usage data
- An industry developed framework
- Outlines high level concepts and principles to be addressed; companies provide specificity
- Respects existing laws, regulations, governance policies, and business environments



- 1. Notice & Awareness
- 2. Choice & Consent
- 3. Data Access
- 4. Integrity & Security
- 5. Self Enforcement &
 - Management & Redress

DataGuard = Consumer Assurance + Responsible Sharing





DOE Program Activities



Recruit Industry Adopters

Awareness campaign focused on third party vendors

Partner Program for Member Organizations

 Mechanism for member-based organizations to indicate their support

Technical Assistance to States

- Assist commissions considering new/updated data access, sharing and privacy regulations
- VCC as a framework







Backup Slides



