

# PMU ANALYTICS

Big Data Analytics Platforms Architecture Requirements and Analysis Techniques

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NASPI Work Group Meeting

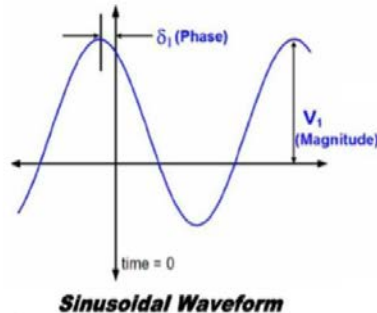
San Diego, CA

April 15-17, 2019

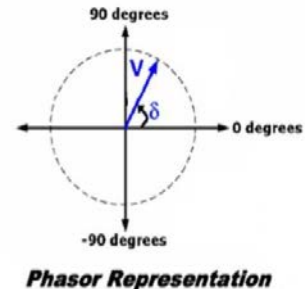


# Challenges of Phasor Data

- Phasor Measurement Unit, aka Synchrophasor
- Measures A, B and C phase currents and voltages and uses a recursive algorithm to calculate the symmetrical components via the Discrete Fourier Transform
- Records 30-120 samples per second (SCADA typically samples once every 2-4 seconds)
- Provides local phase angle and frequency measurement (SCADA measures magnitude and estimates phase angle)
- Time synchronization of PMUs via GPS aligns phase angles to a common time reference allowing for observability across a wide area



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# Challenges of Phasor Data

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## Streaming Data

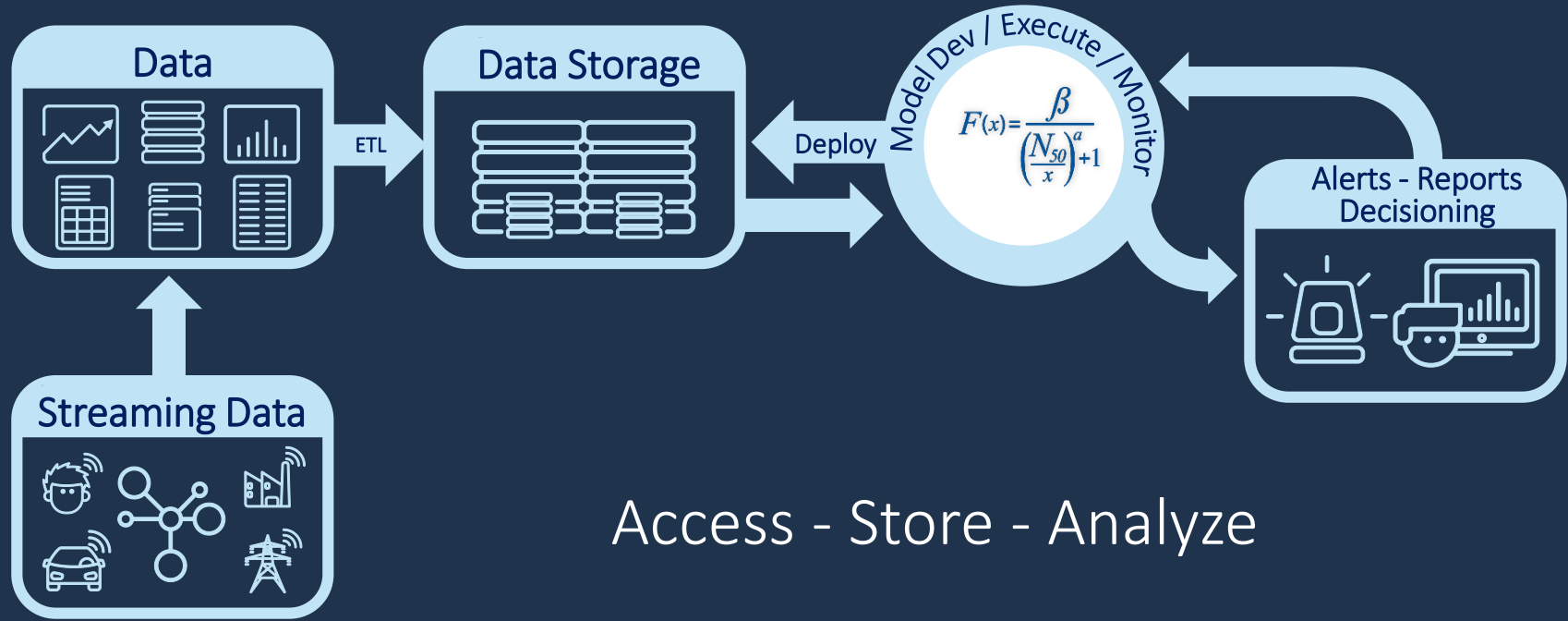
- High speed, real-time
- Continuous analysis
- Specific historical context



## Big Data

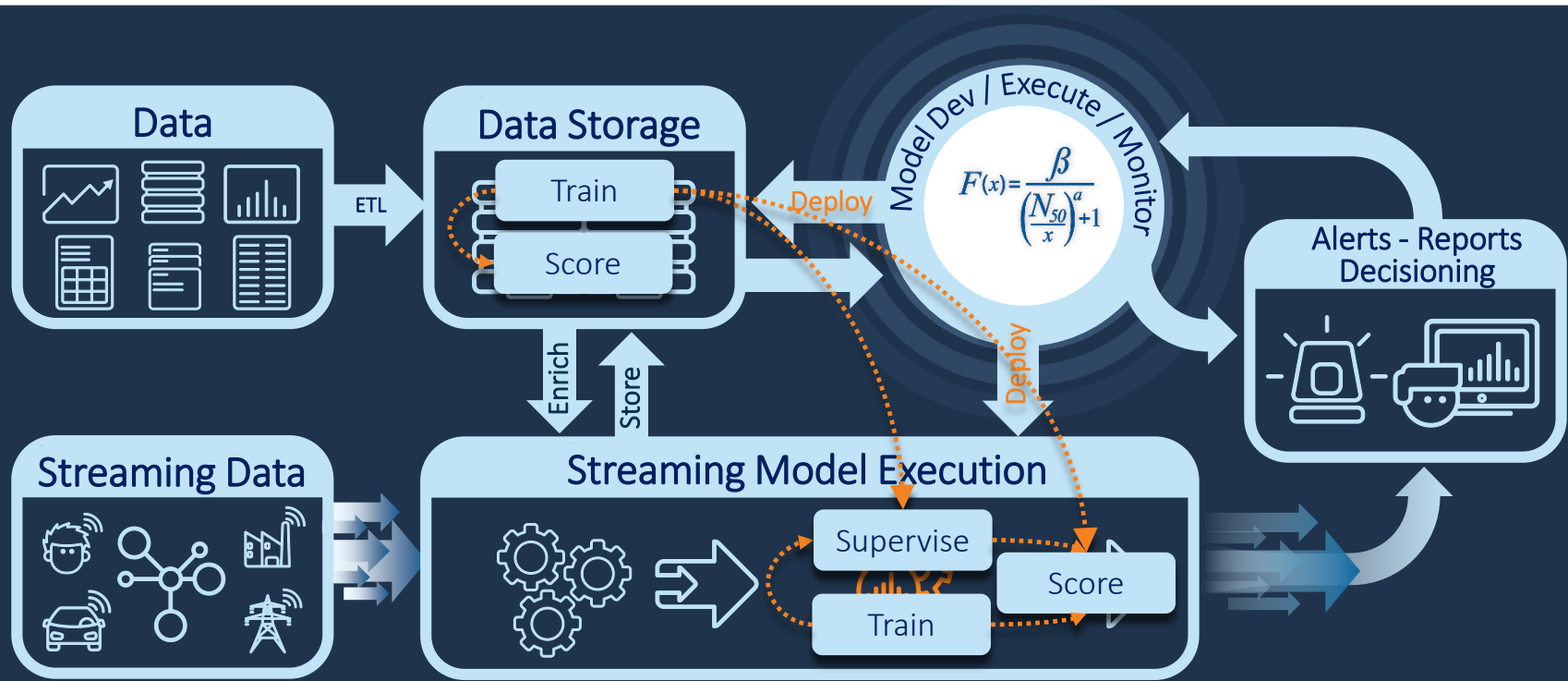
- Large volume
- On-demand analysis
- Full historical context

# Traditional Approach to Data Analysis

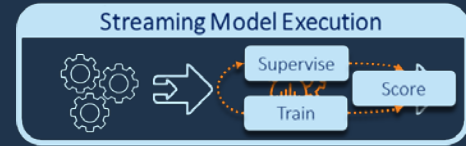
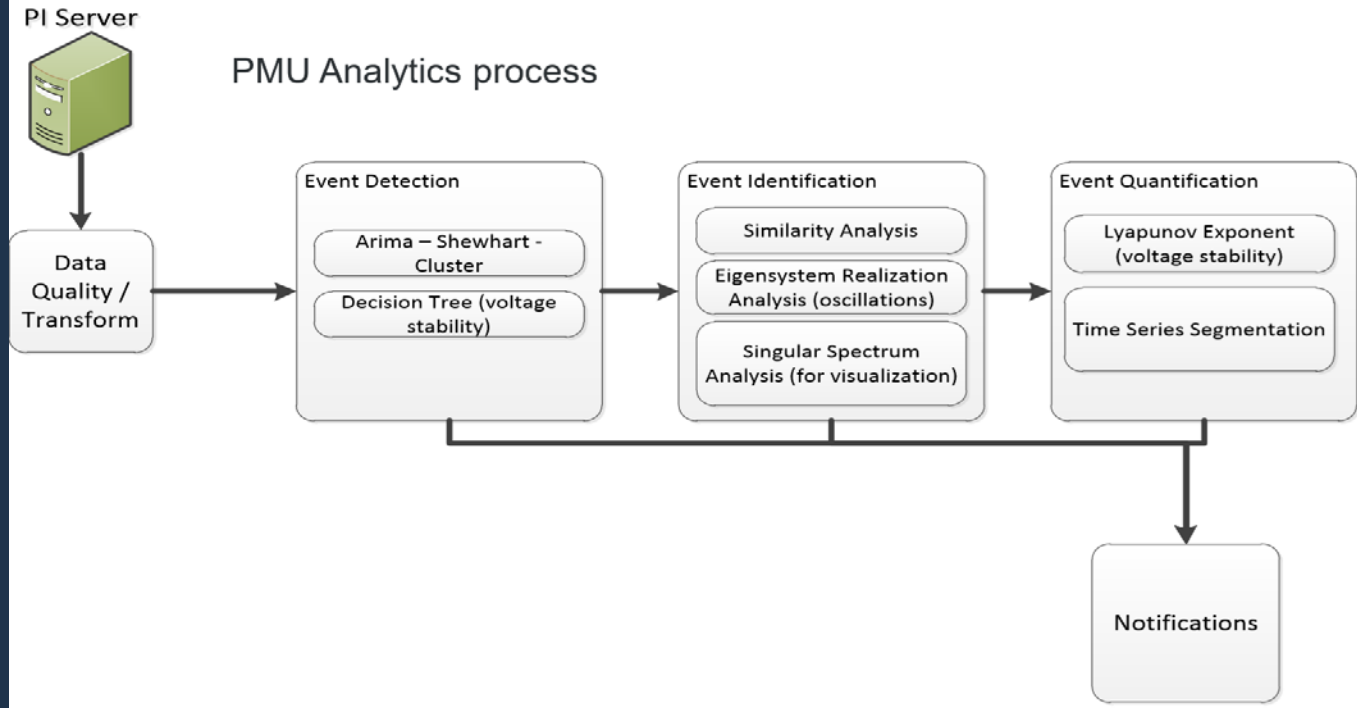


# SAS<sup>®</sup> Approach to Data Analysis

## Stream – Understand – Act



# Analytics In PMU Time



- Data Quality
- Event Detection
- Event Identification
- Event Quantification

# Analytics In PMU Time

## Data Quality

### Data Preparation and Quality

- Working with PI System compressed data
- Use uncompressed tag (Frequency) to detect missing time periods
- **Difference between missing data and bad data**
- **Monitor status tag.** Status may be updated at multiple points in the data chain (At the PMU, at the data collector, in the PI System)
  - Sometimes status in the measurement field
- **Cross check values for consistency**
  - Freq = 0, Angle = 45 (loss of GPS signal)
  - Some PMUs were configured differently (Freq tag), this was corrected
  - Missing measurement during phase angle “wrap” – do not interpolate
- Calculate phase angle differences between PMU pairs
- Automate data preparations and analytics



- **Data Quality**
- **Event Detection**
- **Event Identification**
- **Event Quantification**

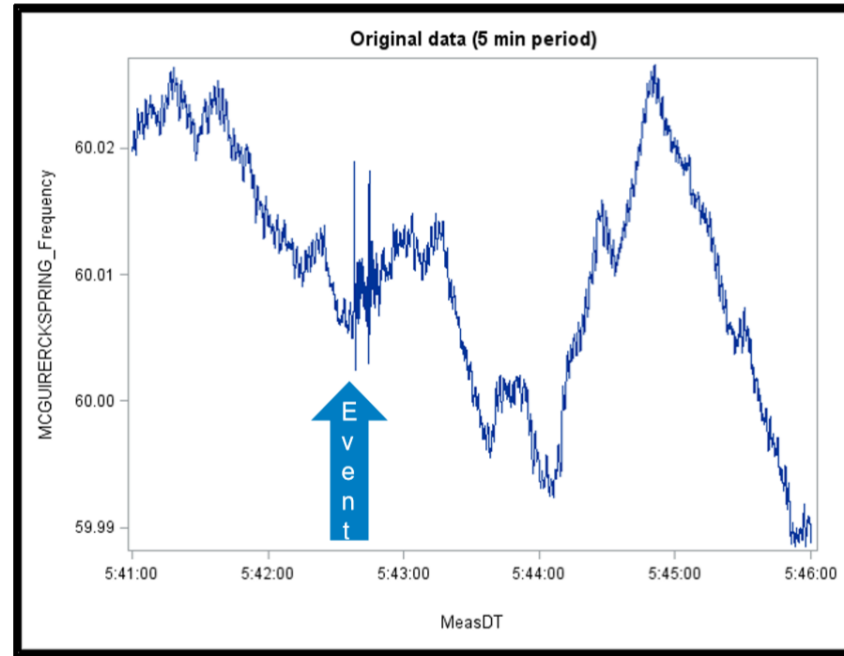
# Analytics In PMU Time

## Event Detection

### Event Detection

**Problem: how to detect events that occur within specs**

- Frequency varies within engineering specifications
- Events occur, but are still within specification



- Data Quality
- **Event Detection**
- Event Identification
- Event Quantification



# Analytics In PMU Time Event Detection

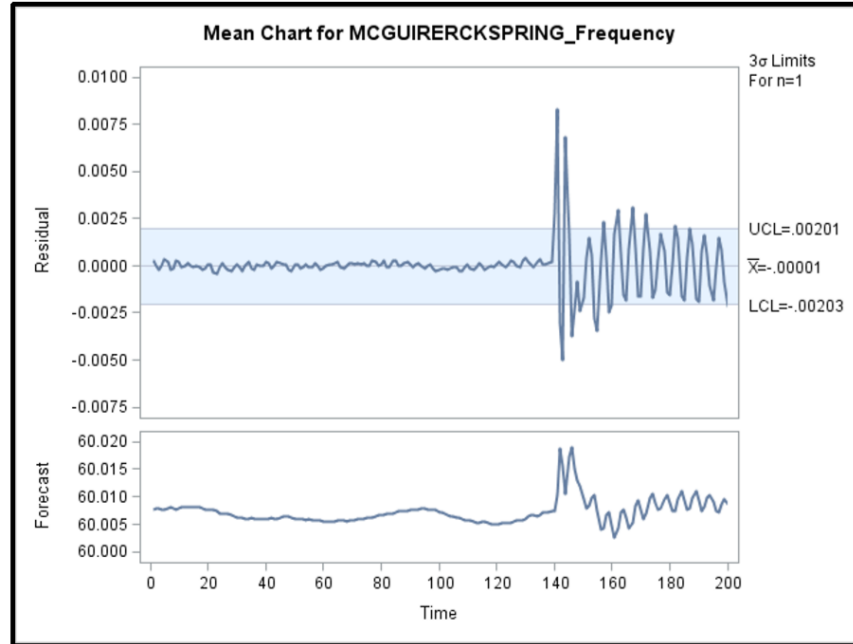
## Event Detection

**Solution: Forecast  
expected values and  
detect deviations**

Residual – difference  
from expected value



Expected value  
based on times series  
model



## Streaming Model Execution



- Data Quality
- **Event Detection**
- Event Identification
- Event Quantification

# Analytics In PMU Time Event Detection

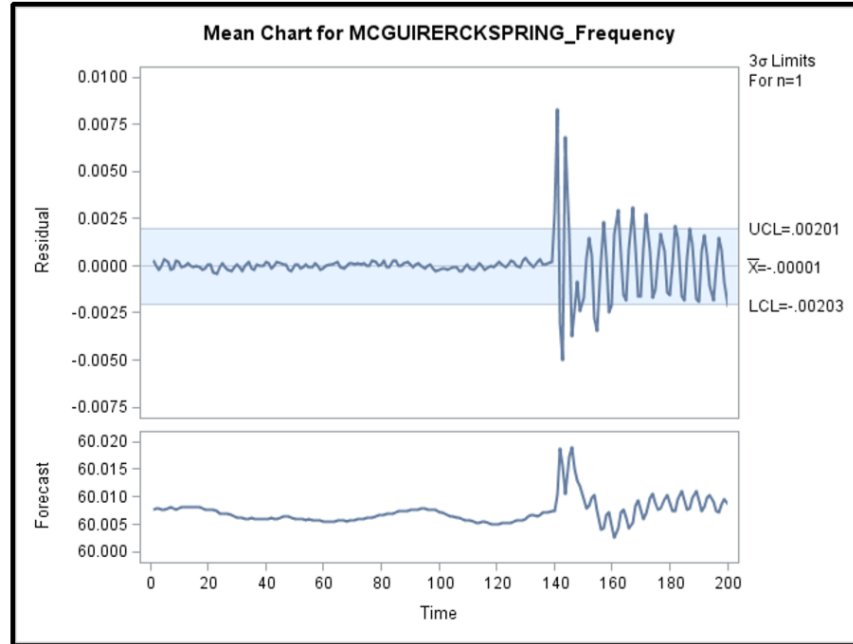
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## Streaming Model Execution



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- **Event Detection**
- Event Identification
- Event Quantification

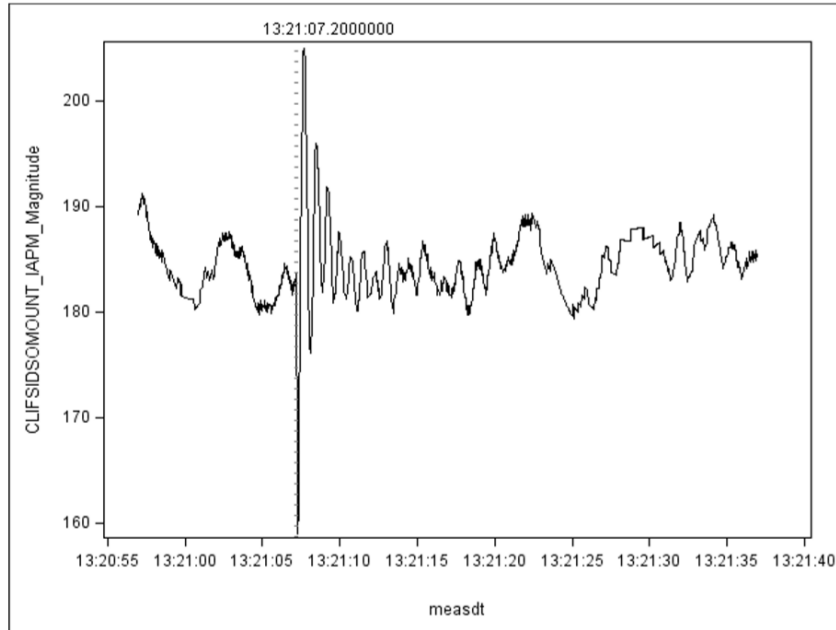
# Analytics In PMU Time

## Event Identification

### Event Identification

**Problem: How to take incoming events and categorize them**

- Current oscillates after event, but then dampens down to normal



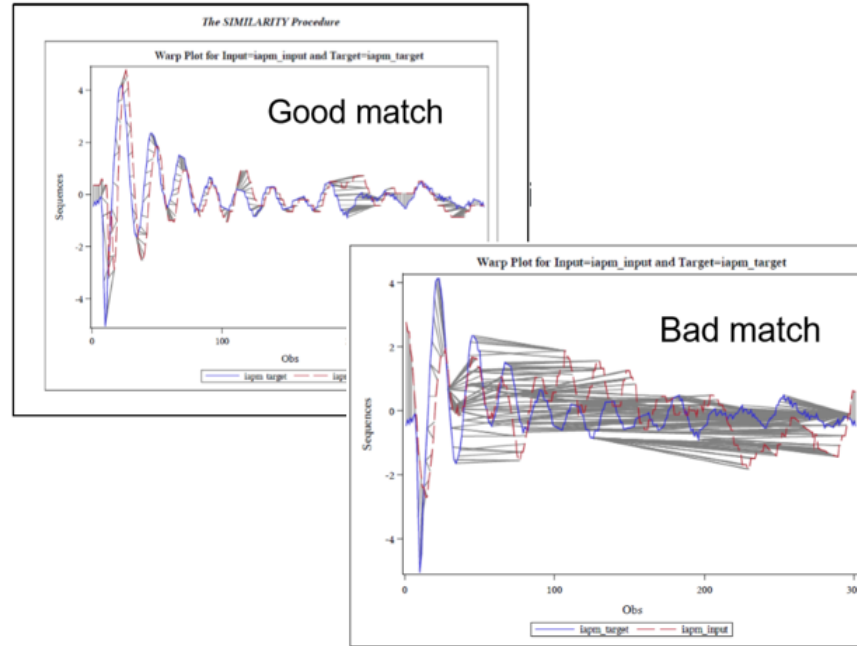
- Data Quality
- Event Detection
- **Event Identification**
- Event Quantification

# Analytics In PMU Time Event Identification

## Event Identification

**Solution: Use  
similarity analysis and  
time-series data  
mining to categorize  
data streams**

- Similarity between incoming stream and reference time series are measured and quantified



## Streaming Model Execution



- Data Quality
- Event Detection
- **Event Identification**
- Event Quantification

# Analytics In PMU Time

## Event Quantification

### Problem: How to predict events which are rare

- situations that are vulnerable to voltage collapse



### Solution: Learning loop system using simulation data to create predictive models

- Use PSS/E simulation software to generate cases for voltage stability
- Build decision tree and use Lyapunov Exponent to identify vulnerable situations



- Data Quality
- Event Detection
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# Analytics In PMU Time

## Event Quantification

### Decision Tree

- Decision Tree
  - Data mining technique that uses large set of data with a desired object of analysis (target) and creates rules from the data that has the highest prediction of the target value
  - Benefits - simple to use and understand, robust to different data types and quality issues
- Using the Decision Tree to Predict Voltage Collapse
  - PSSE – Simulations
    - Ran 3,128 dynamic simulations of a generation loss event using different combinations of load and operating conditions
    - Collected voltage and power data from system PMU locations
    - Classified each simulation as either Voltage Stable or Voltage Collapse
  - SAS Enterprise Miner – Decision Tree Builder
    - Input snapshot of simulated PMU data prior to generation loss event and the simulation classification
    - Builder partitions data for tree creation, validation and testing
    - Builder finds a given number (4 in this case) of the highest predictors
    - Builder tests the results with the data partitioned for testing



- Data Quality
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- **Event Quantification**

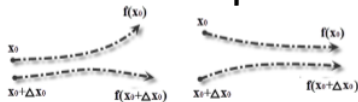
# Analytics In PMU Time

## Event Quantification

### Lyapunov Exponent

- Characterizes the rate of convergence or divergence of non-linear dynamical systems.
- For stable/unstable systems the Lyapunov Exponent will be negative/positive

- General Equation



$$d(f^n(x_0), f^n(x_0 + \Delta x_0)) = e^{\lambda n} d(x_0, x_0 + \Delta x_0)$$

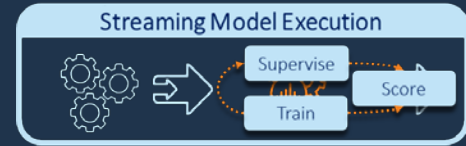
$$\lambda = \frac{1}{n} \ln \left[ \frac{d(f^n(x_0), f^n(x_0 + \Delta x_0))}{\Delta x_0} \right]$$

[reference 1]

- Time Series Calculation

$$\frac{1}{Nk\Delta t} \times \sum_{m=1}^N \log_{10} \frac{|v_{(k+m)\Delta t}^i - v_{(k+m-1)\Delta t}^i|}{|v_{m\Delta t}^i - v_{(m-1)\Delta t}^i|}, k > N$$

[reference 2]

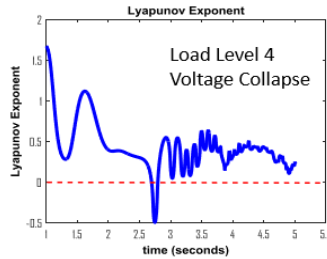
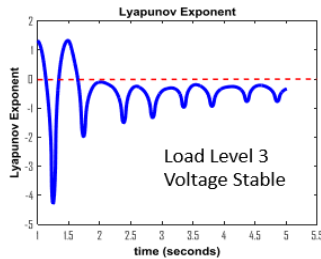
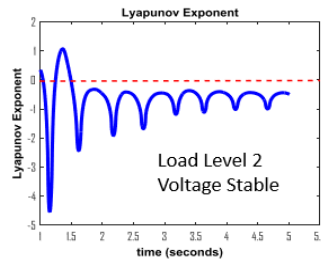
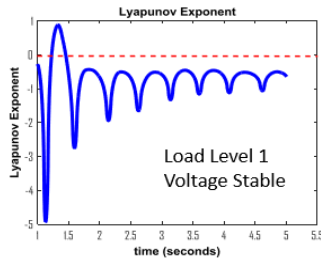


- Data Quality
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# Analytics In PMU Time

## Event Quantification

### Lyapunov Exponent 500 kV Line Trip



- LE data for 4 cases with the same operating conditions but different voltage sensitive area loads over a window of 5 seconds for a 500 kV Line Trip Event
- For the 3 Voltage Stable cases the final value of the LE is negative and the margin between zero (red dashed line) and LE (blue) decreases as load in voltage sensitive area increases
- For the Voltage Collapse case the final value of the LE is positive

#### Streaming Model Execution



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# Analytics In PMU Time

## Event Quantification

### Future Work

- Event Classification:
  - Develop a library of event signatures from PMU data to compare to incoming PMU data.
- Decision Tree:
  - Scan PMU data in real time for current prediction of voltage collapse decision tree rules.
  - Build Decision Tree for more common events and test the results on PMU data in real-time.
  - Create several prediction of voltage collapse decision trees, one for each contingency, and apply the results to PMU data in real-time.
- Lyapunov Exponent:
  - See if there is a faster way to determine the final sign of the LE and take control actions locally in time to prevent voltage collapse.



- Data Quality
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