

# Hunting for Anomalies in PMU Data

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# “Big Picture” Objective



Power grid related data  
(PMUs, State  
Estimators, Load, etc)

Analytical Tool that provides:

- Real time analytics, monitoring the state of the grid
- Capability to look at historical trends and events
- Reliable predictions about the forthcoming state of the grid

# Pre-Processing Steps

- Read and store raw **PMU data**

- Apply a data quality filter and calculate signatures (features) for analysis



**State Estimator Data**

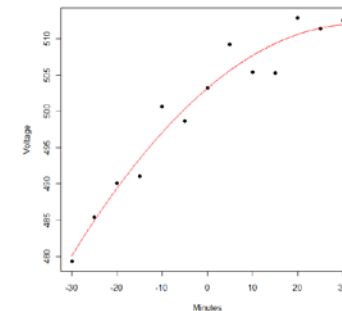
1 Yr = 5 GB

**PMU Data**

1 Yr (45 PMUs) = 3 TB



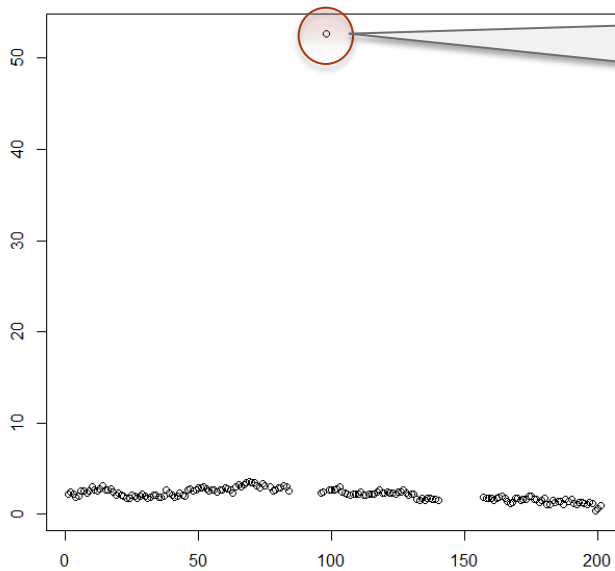
Signatures capture **mean** and **variability** as well as temporal aspects like **slope** and **acceleration**



# Data Storage

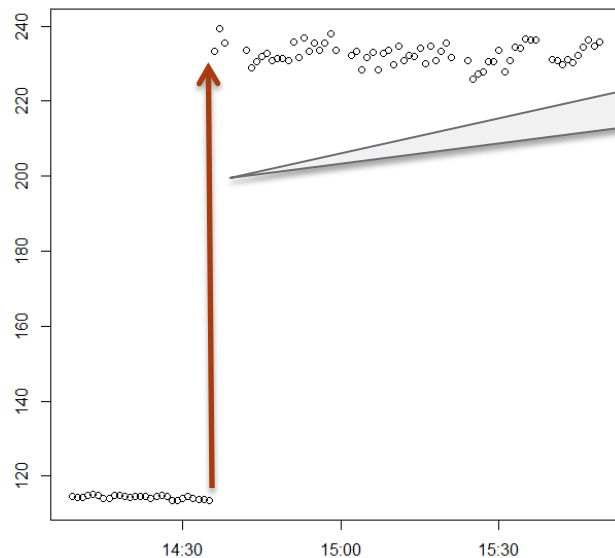
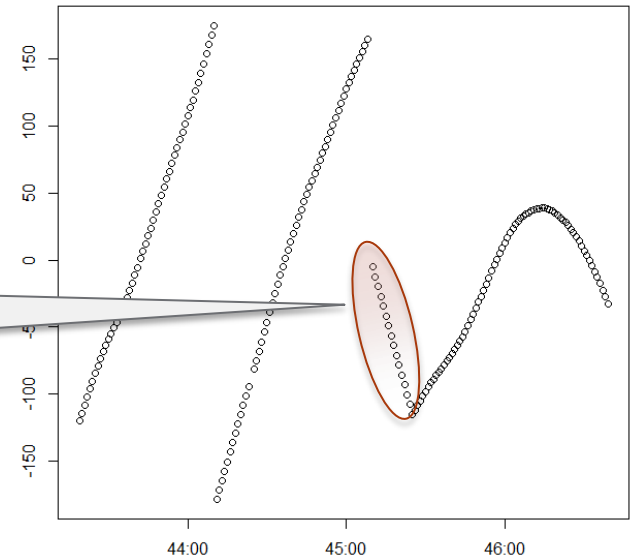
- ▶ **2 Months** of Eastern Interconnect PMU data, down-sampled to 1 per second. Focus was studying phase angle pairs.
- ▶ Nearly **2 years** of BPA PMU data, measured 60 times per second. Focus on frequency, voltage, current, & phase angle pairs.
- ▶ Data were **not** stored in a **Hadoop** cluster or any databases. They are difficult to set up and can be difficult to input data or export data.
- ▶ Data were stored in small, 1 minute files in Rdata format. This enabled the data to be easily read into R or Matlab for any additional plotting or analyses.
- ▶ This shifted the burden of data management on us, instead of relying on the software and the Hadoop or database experts.

# Data Quality Filters



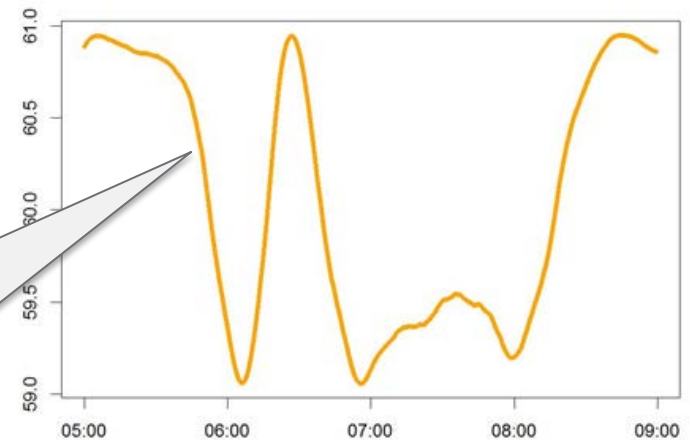
Usually easy to deal with

Phase angle wrap around issue

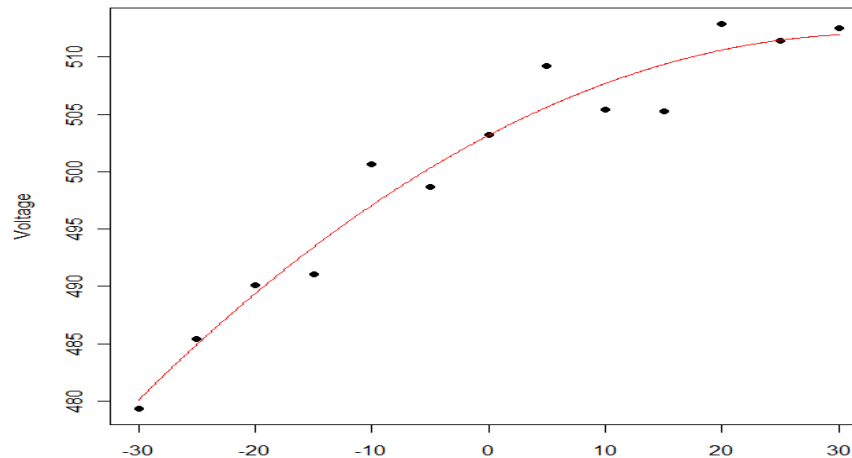


Sudden changes in phase angle

Impossible variable behavior (frequency ranging from 59 to 61)



# Feature Extraction (Data Signatures)



- ▶ Regression fits through the data calculate estimates of **value**, **slope**, **curvature** (acceleration), and **noise**.
- ▶ Can be calculated in the presence of missing or data quality flagged values.
- ▶ Summaries of these features are used in the analyses.

► Find **patterns** in the data

Apply **clustering** algorithms to determine patterns (similarities) in the data.

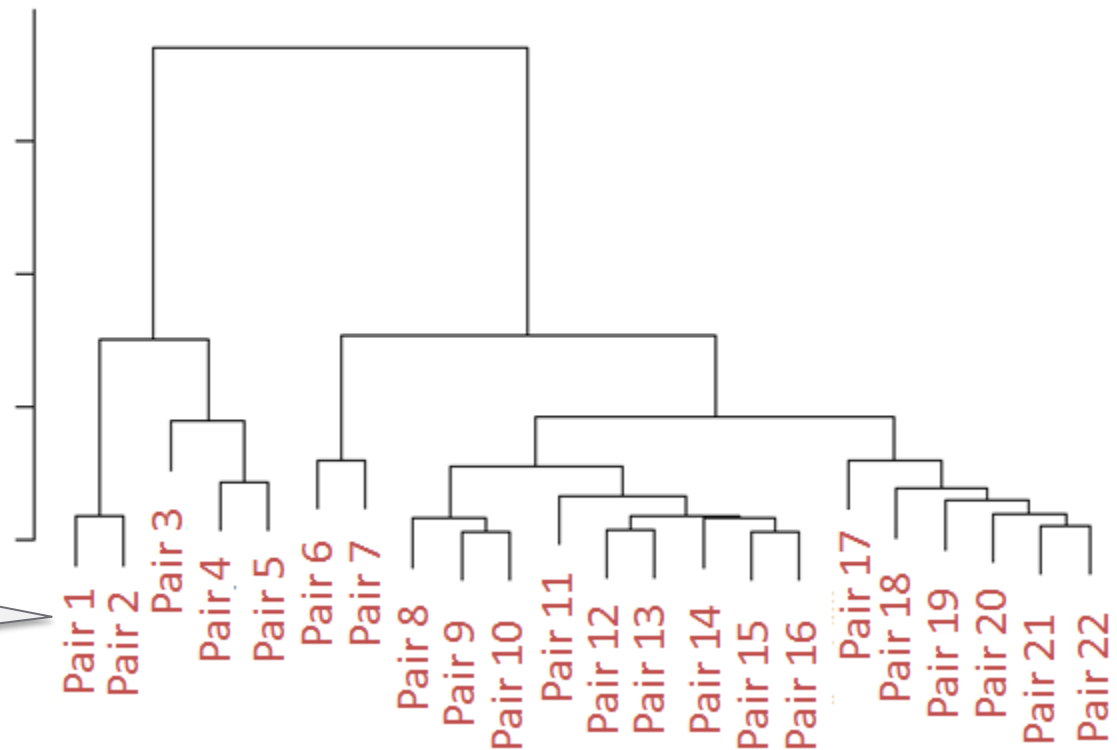
These **similarities** could be determined by **time** (i.e. periods of time in which the grid is in a similar state) or by **variables** (i.e. variables that are closely related).

► Find **anomalies** in the data (referred to as **atypicalities**)

Apply **multivariate** analytical techniques to find **unusual** moments in time with respect to a **baseline** that was determined from **historical data**.

# Eastern Interconnect Clustering of Phase Angle Pairs

**Cluster Dendrogram**



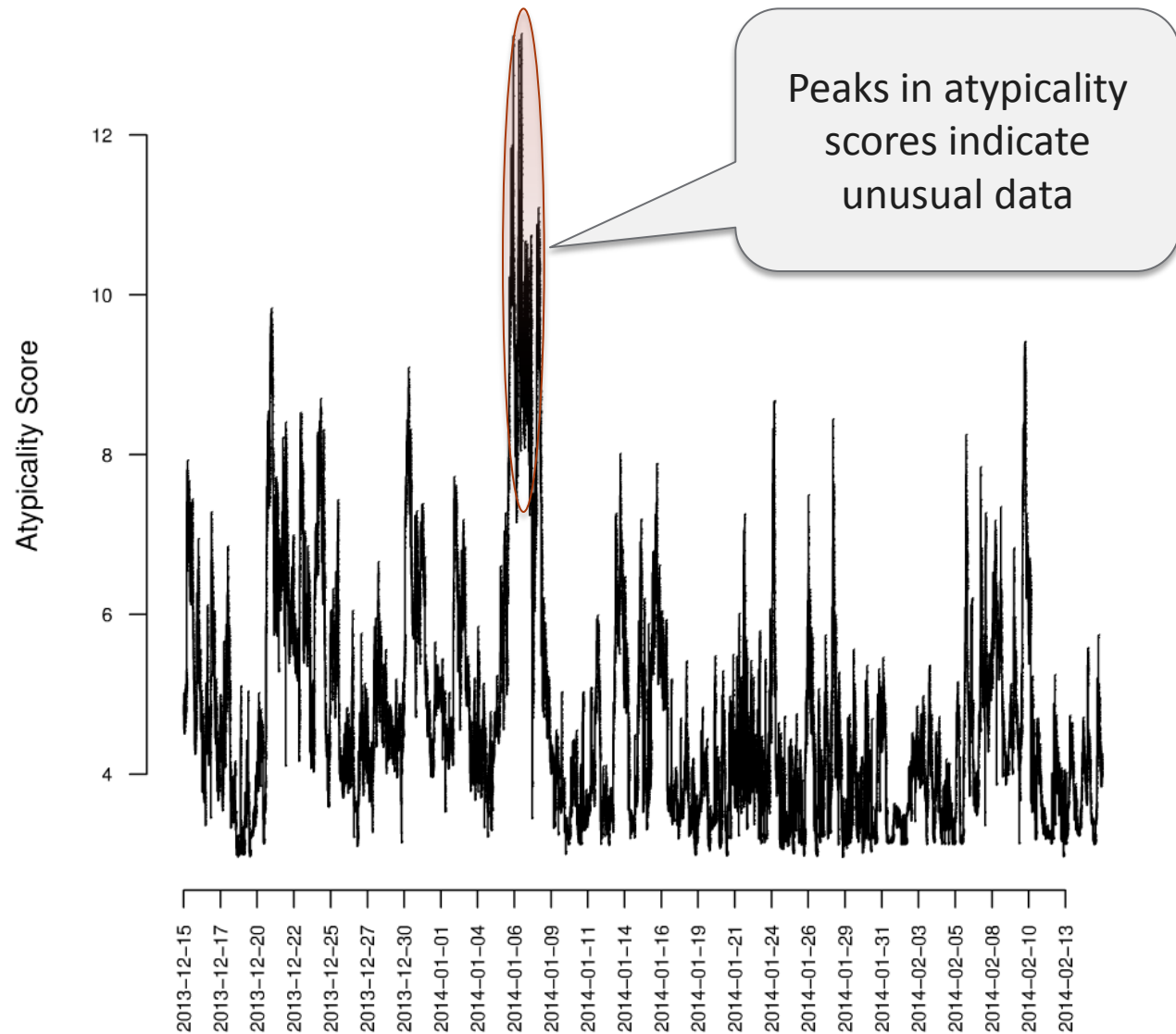
Proximity on tree  
indicates similarity



# Eastern Interconnect PMU Data Atypicality Detection

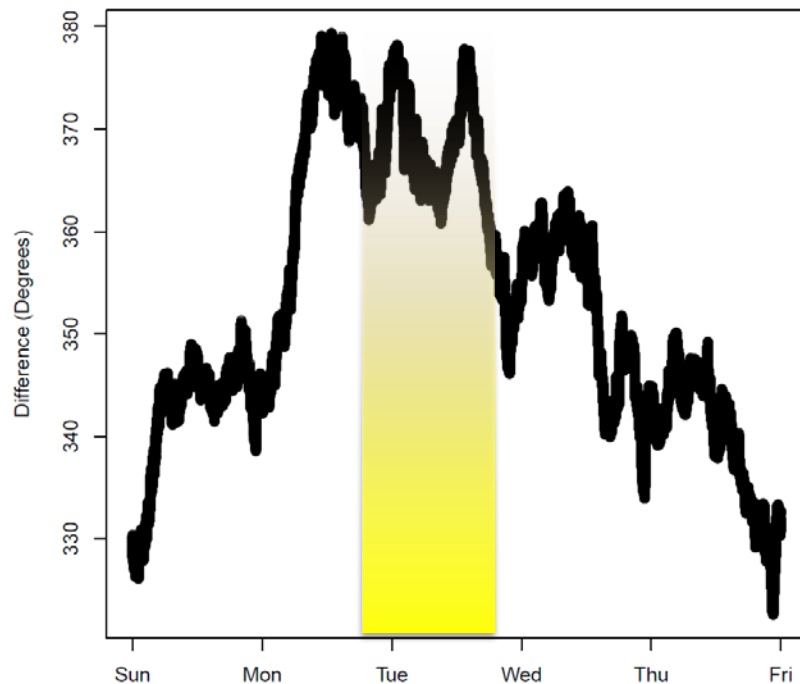
Analyses performed  
on selected phase  
angle pairs across  
the Eastern  
Interconnect

**This anomaly  
corresponded to a heavy  
load time period**

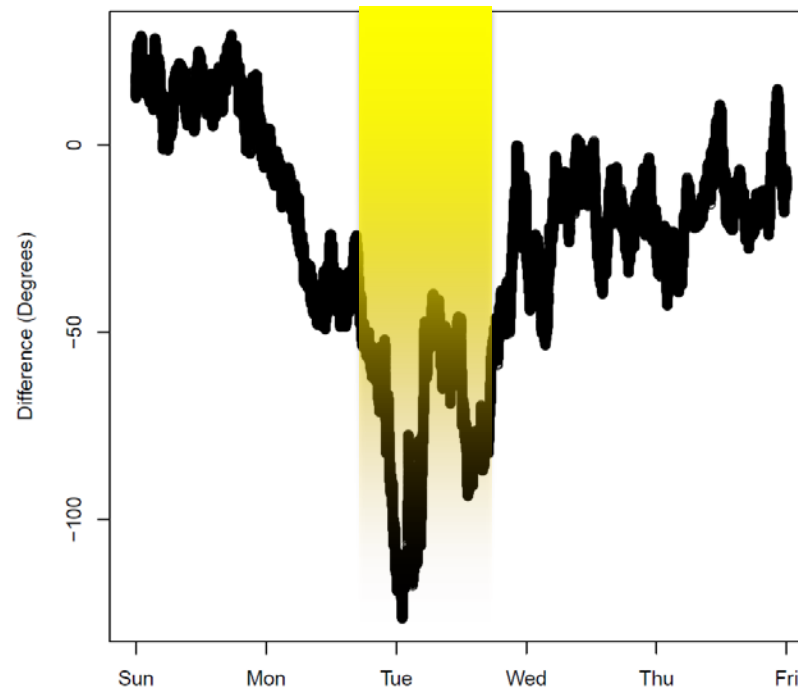


# Eastern Interconnect Drilldown Plots

**Pair 2**



**Pair 17**

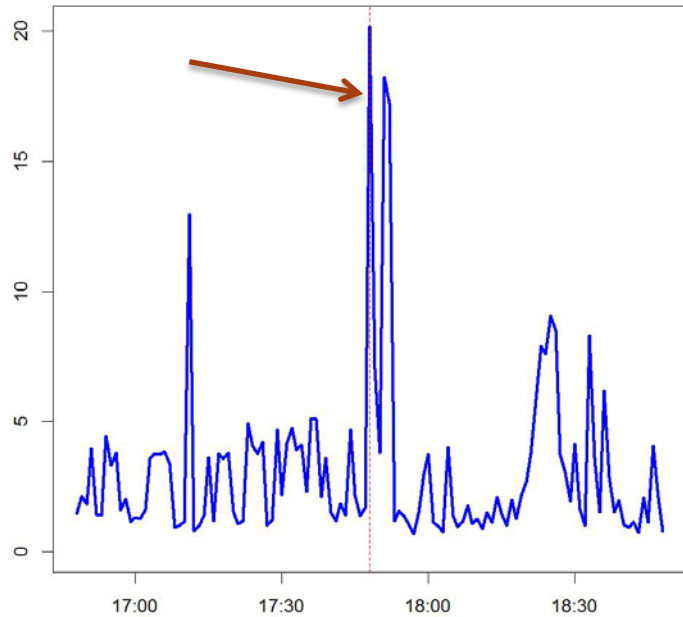


Two angle pairs that showed atypical behavior when the atypicality score was high. Pair 2 shows atypical behavior before the atypicality score triggered.

Studying this type of behavior may help discover possible precursors to atypical events.

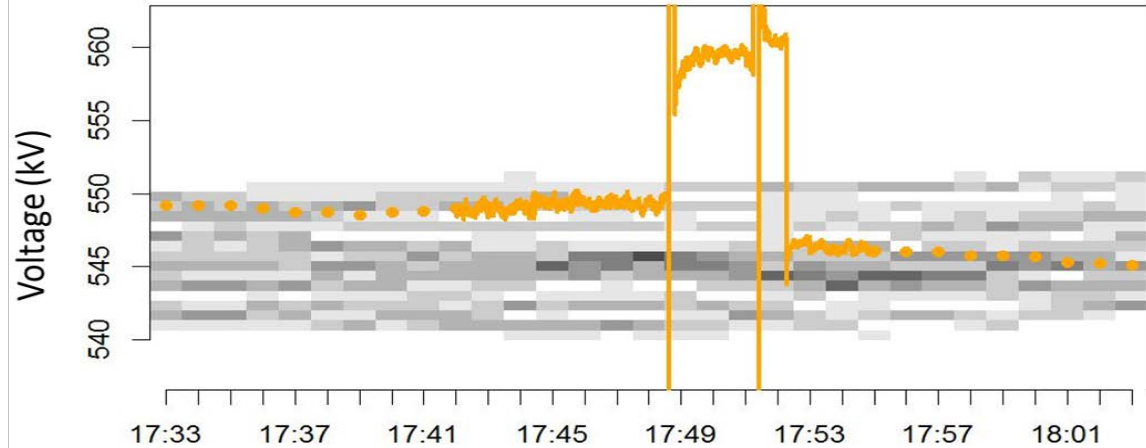
# BPA PMU Data Atypicality Detection

## Atypicality Score

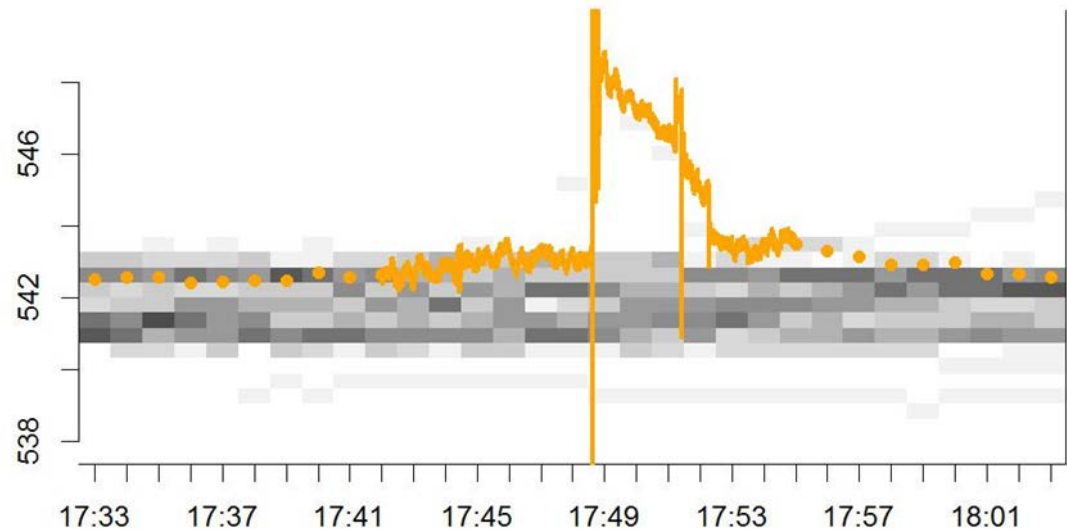


## Lightning related anomaly

## Substation A

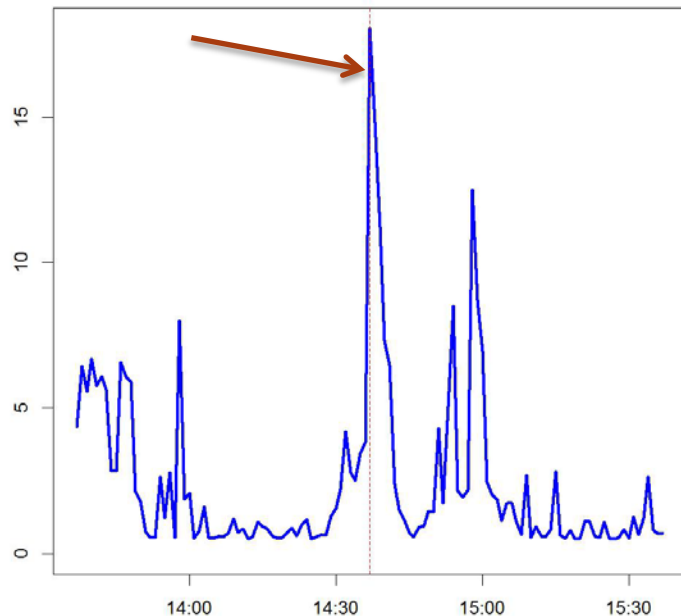


## Substation B

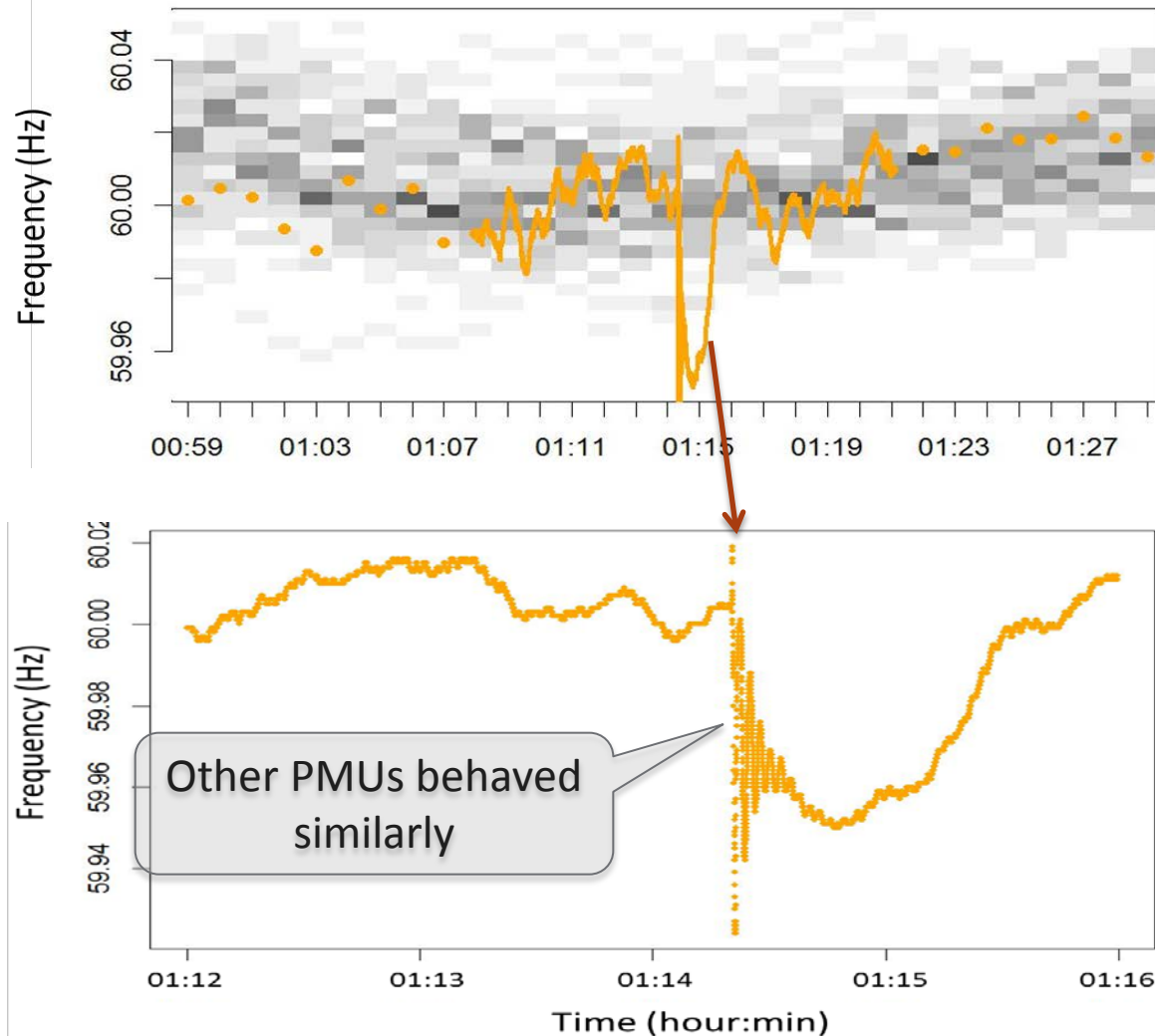


# Atypicality Detection - Example

## Atypicality Score

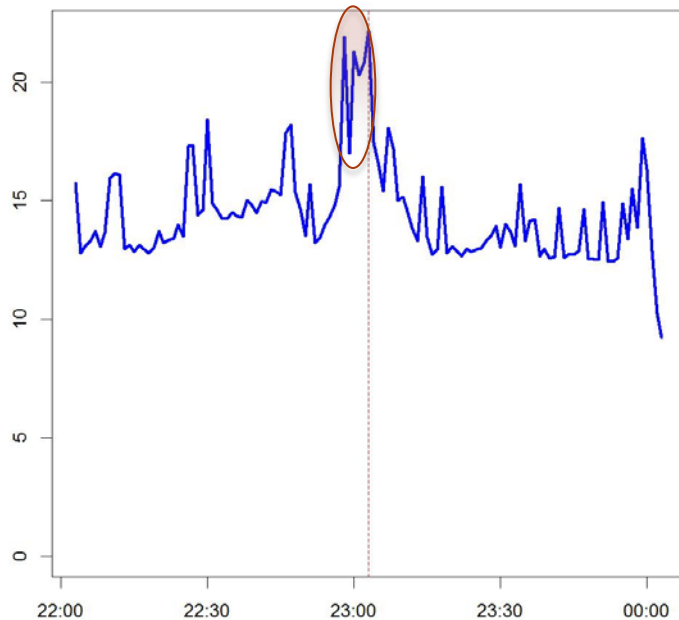


**Equipment failure related anomaly**

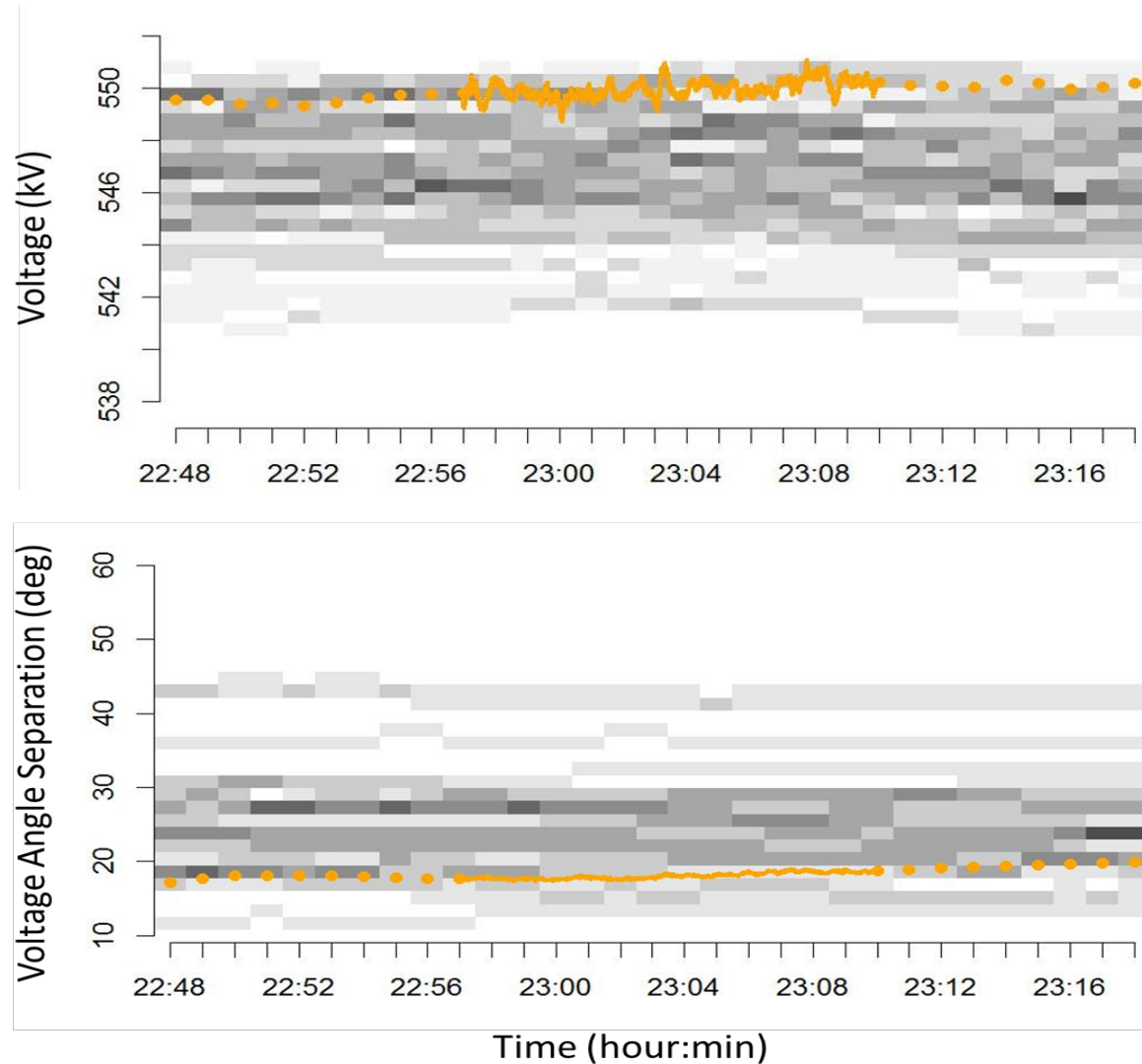


# Atypicality Detection - Example

**Atypicality Score**



**An example of a detected anomaly that was not related to a specific system event**



# Next Steps

- ▶ Attach these tools to a PMU data stream for near-real time results.
- ▶ Add other data streams for analyses.
- ▶ Process a longer time period.
- ▶ Determine what thresholds should be used to alert when data are atypical.
- ▶ Investigate data for predictive patterns.