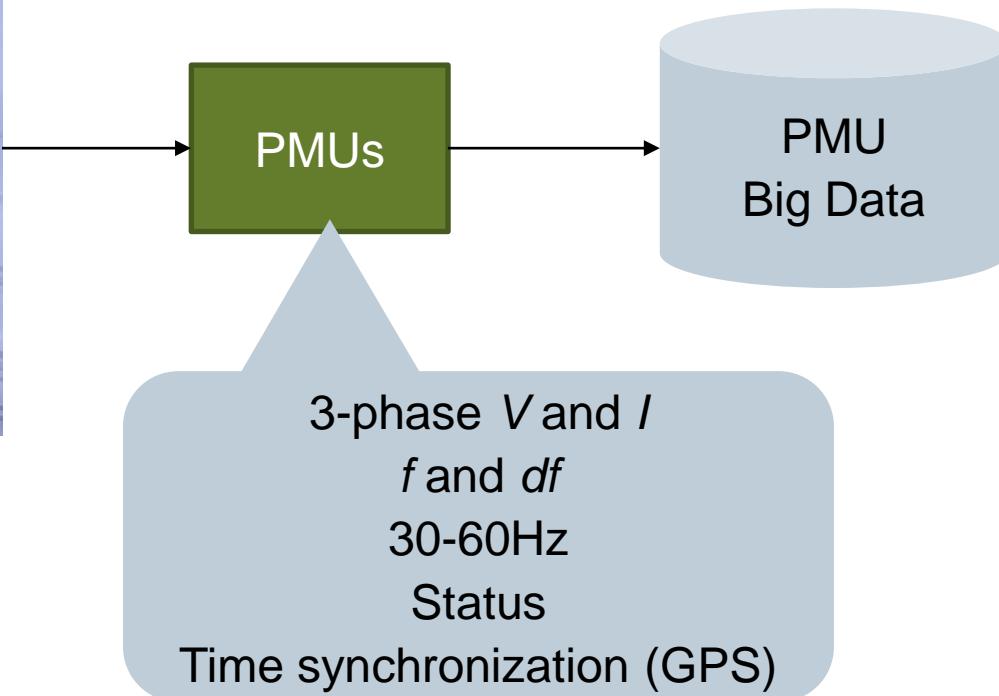


# Big Data Processing for Power Grid Event Detection

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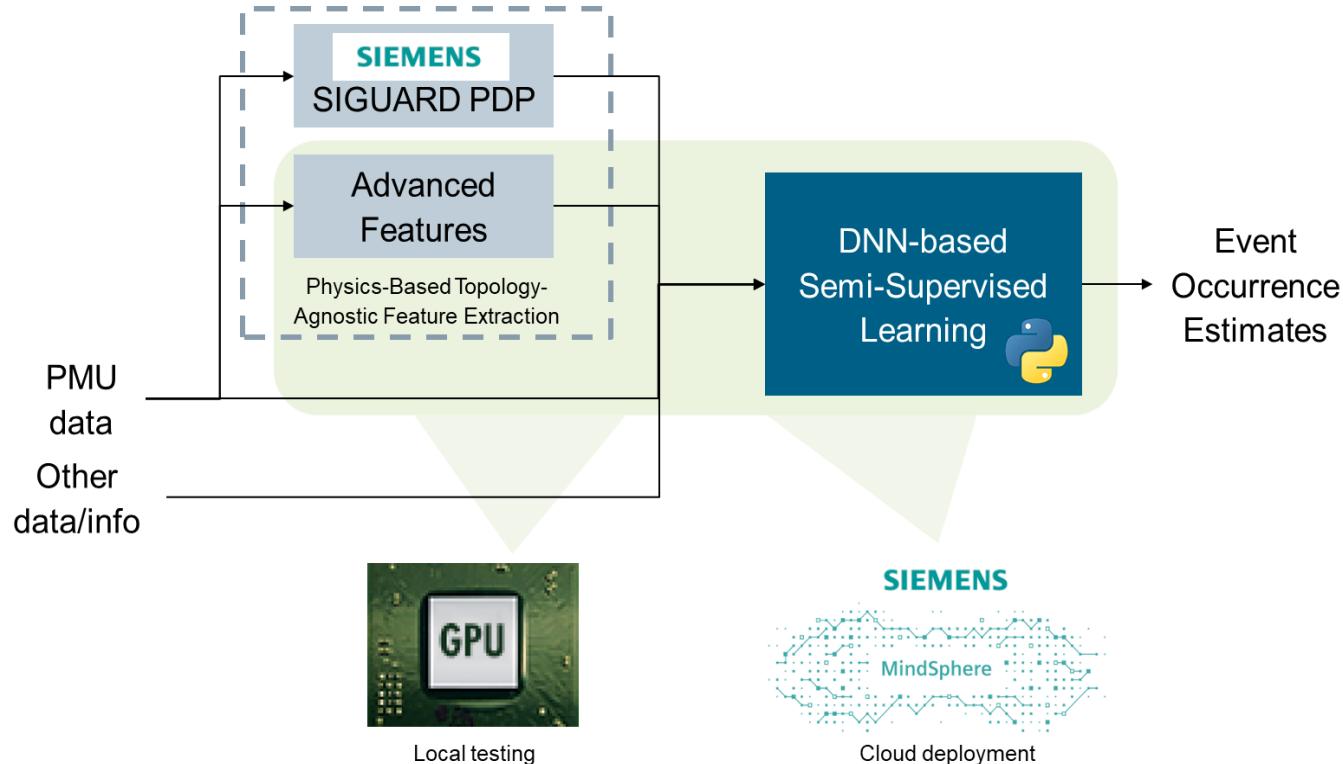
# Introduction



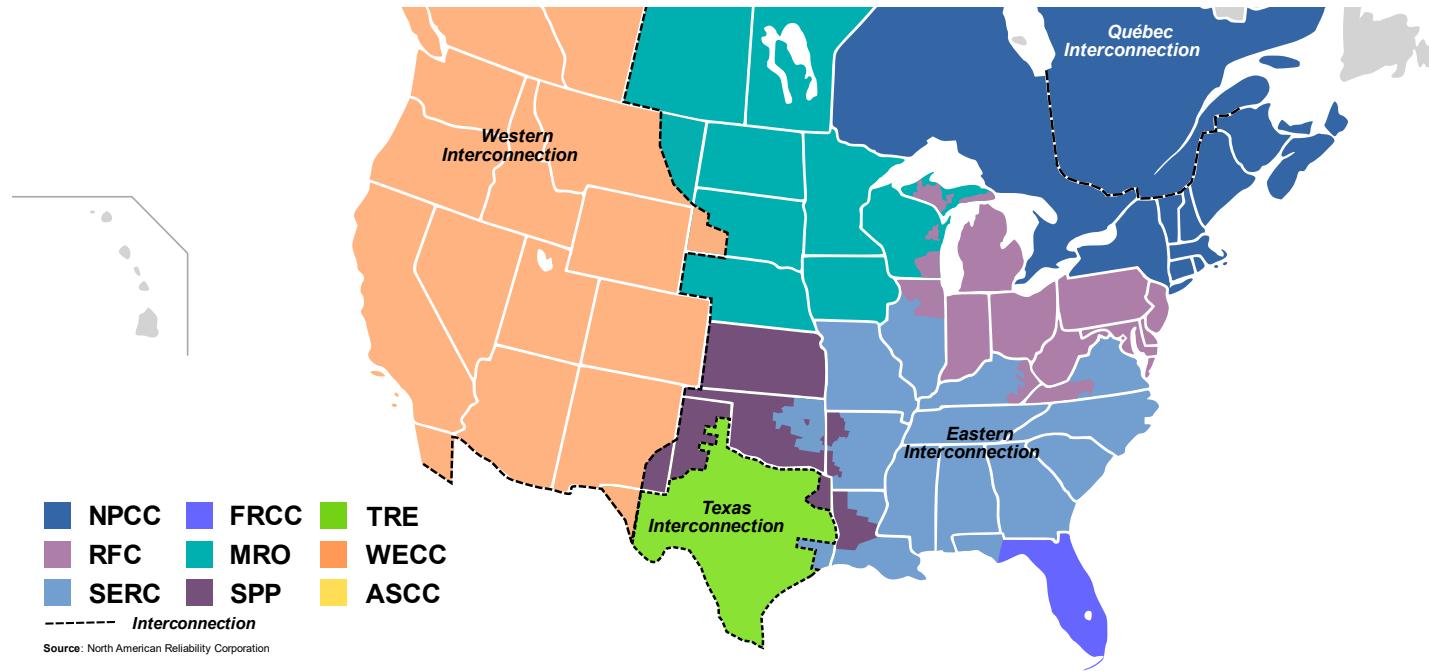
# Introduction



## MindSynchro: DOE FOA 1861



# Big Data and Processing



Interconnection	A	B	C
Actual IC	Texas IC	Western IC	Eastern IC
Start Date	2018-07-21	2016-01-01	2016-01-01
End Date	2019-08-24	2017-12-31	2017-12-31
PMU Number	215	43	188
Data Volume	~3TB	~5TB	~11.5TB
File Count	2576	4365	10496

## Challenges about the Data

### PMU Data quality:

- Missing values or missing complete measurements
- Duplicate rows
- Unaligned timestamps for different PMUs
- Overlap between files

### “Label” information: Event logs

- Purpose of the log is not appropriate for labeling data
  - Label doesn't reflect the underlying phenomena but the factor causing the event.
    - E.g. Categories “wind” and “animal” may both refer to short circuit or line down events with same/similar pattern
    - Duration may reflect the consequence of the event (e.g. blackout)
  - No mapping to specific PMUs
  - Manually defined and manually post-processed
  - Large subset of events which are not of interest (e.g. planned events)

## Computational Infra-structure

### GPU Server:

- Processor: Intel® Xeon® Silver 4210, with 40 cores
- 196GB of RAM
- 4 NVIDIA Quadro RTX 6000 GPUs, each with 24GB of RAM
- 2TB NVMe drive for operating system
- 42TB of HDD space for data storage
- Ubuntu 18.04.2 LTS operating system

## Pre-processing

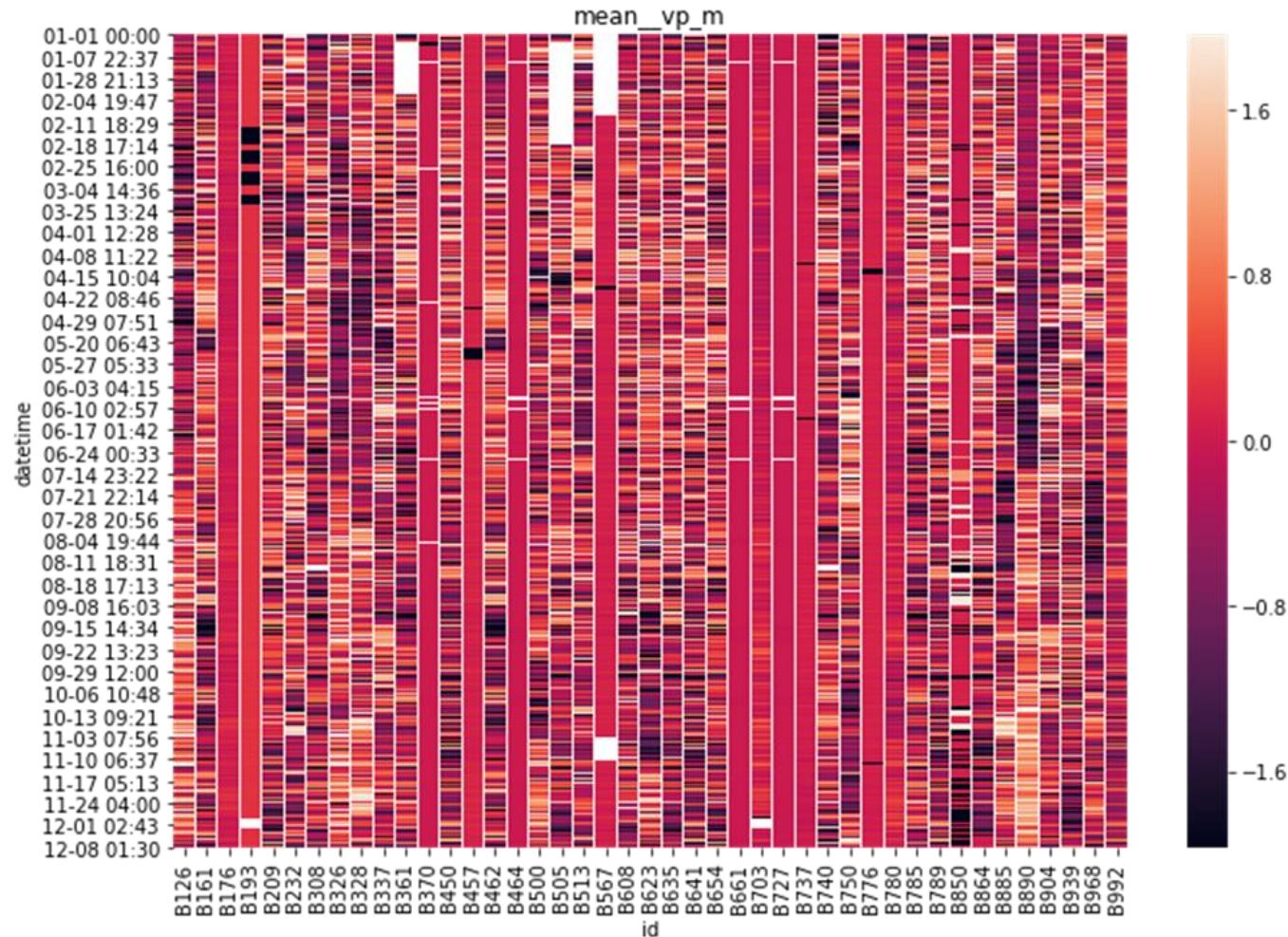
1. loading original parquet file
2. converting numerical fields type (originally stored as strings)
3. converting empty values to null
4. removing columns which are all null
5. converting timestamp type (originally stored as strings)
6. indexing the data based on the timestamp
7. **calculating and saving statistics**
8. **synchronizing the data from all PMUs for each timestamp**
9. sorting by timestamp
10. saving the pre-processed data as a new parquet file

Status and outliers based data masks calculated as a separate pre-processing task

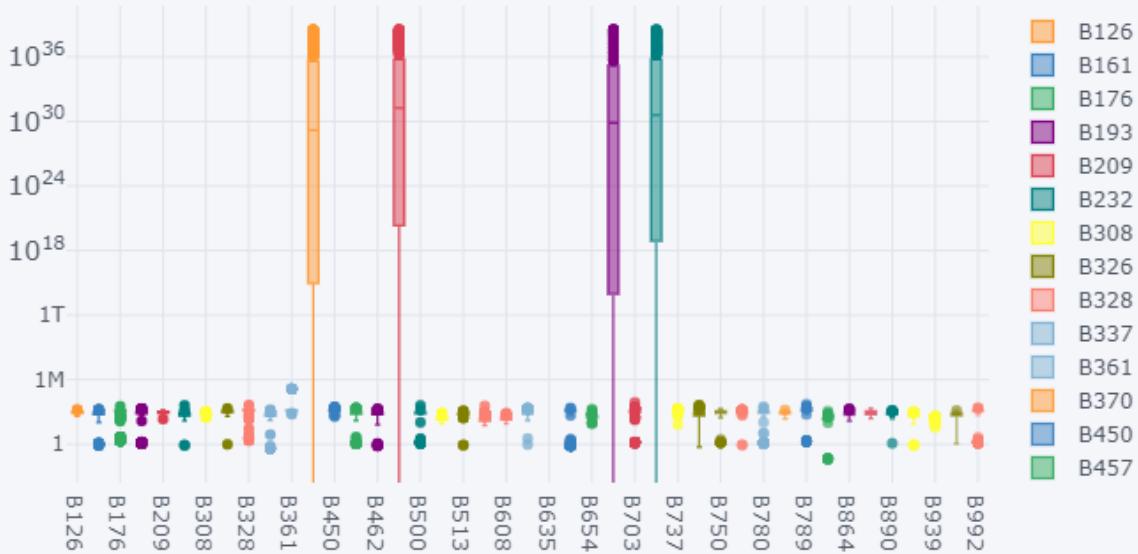
## Leveraging GPU for pre-processing (RAPIDS cuDF)

# Data Visualization and Annotation

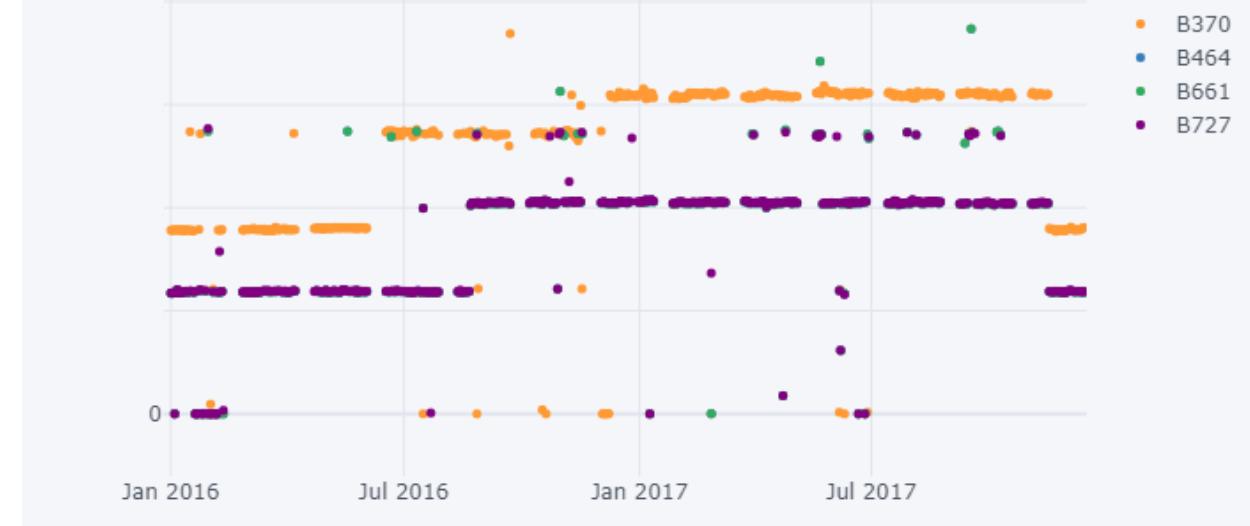
**SIEMENS**  
Ingenuity for life



Boxplot: max\_ip\_m



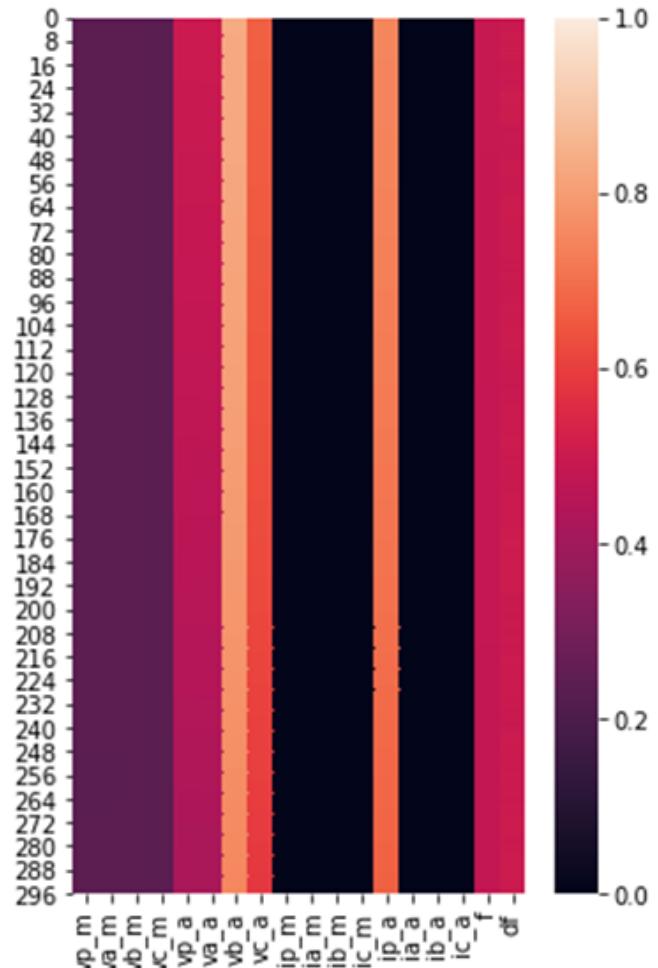
Scatter: median\_vp\_m

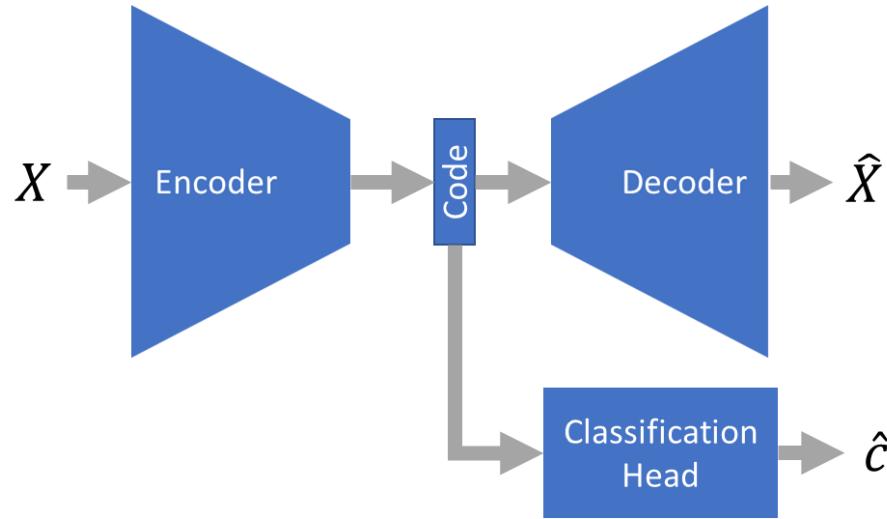




## Batch preparation

1. load parquet metadata from pre-processed file and corresponding valid data mask
2. filter columns to specific PMUs if needed
3. load corresponding columns for data and mask
4. apply valid data mask to the data
5. eliminate regions of overlap with other files
- 6. unwrap phase angles (optional)**
7. fill missing columns with token value
- 8. resample to 30Hz**
- 9. forward fill and then back fill missing values**
10. eliminate extra rows that won't fit in batch
- 11. normalize values based on pre-defined ranges for each measurement**
12. reshape data to batch format
- 13. adjust all angle values to the same reference in each batch (optional)**
14. replace token value from missing columns with zeros





- Tensorflow 2
- Unsupervised learning dataset: over 370k samples (24 PMUs, 2 years of operation)
- Supervised learning dataset: short circuit, 221 labels from which 29 are positive
- Results: perfect accuracy

- PMU provides very rich big data
  - Well suited for detecting and categorizing relevant events
- Proper labels currently not available
- Challenges in dealing with real world big data
- Challenges in using the data for ML
- Sample ML results
- Next Steps:
  - Work in progress (project end expected by mid-2021)
  - Improvement in labels
  - Improvement in ML models
  - Insights from the complete dataset

# Contact page



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