

UCR

Situational Awareness in Distribution Grid Using Micro-PMU Data

(Tasks 1.1 and 1.2)

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Agenda

- **Situational Awareness Using Distribution Synchrophasors**
 - Events in Distribution Systems
 - Event Detection
 - Event Classification

- **Use-Cases:**
 - Asset Monitoring
 - Protection System Diagnosis

- **Cyber Security & Situational Awareness**

Our Focus: Sensors and Measurements

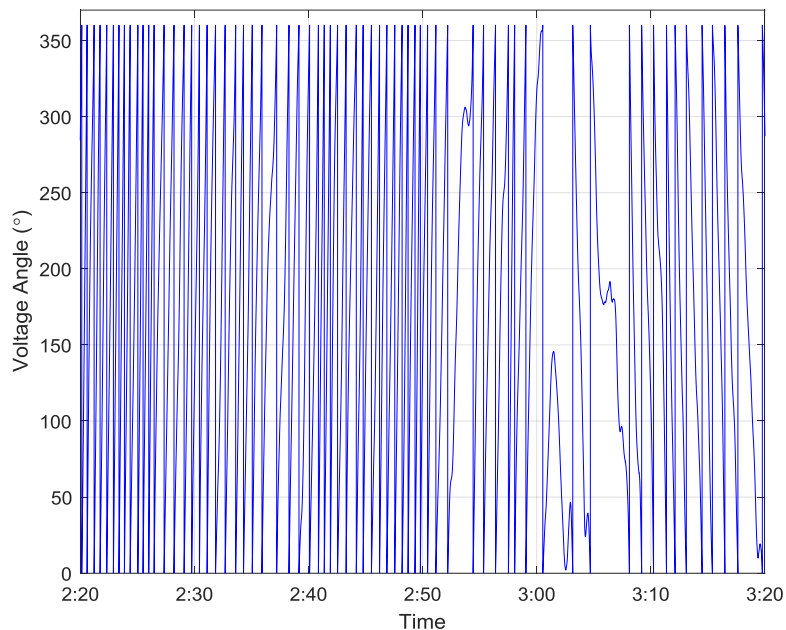
Data Recording:

- Voltage Phasor
- Current Phasor

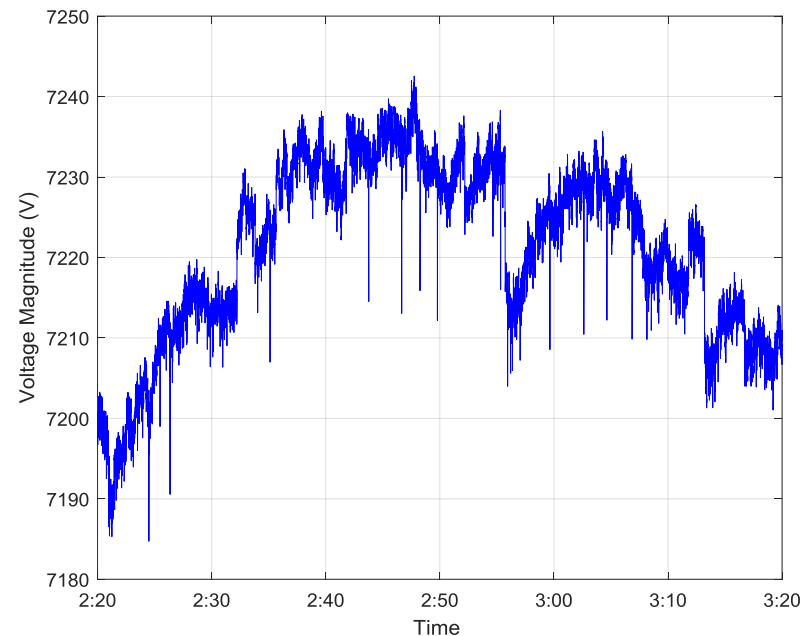


Features:

- Millisecond Reporting (120 Hz)
- 0.1 Accuracy



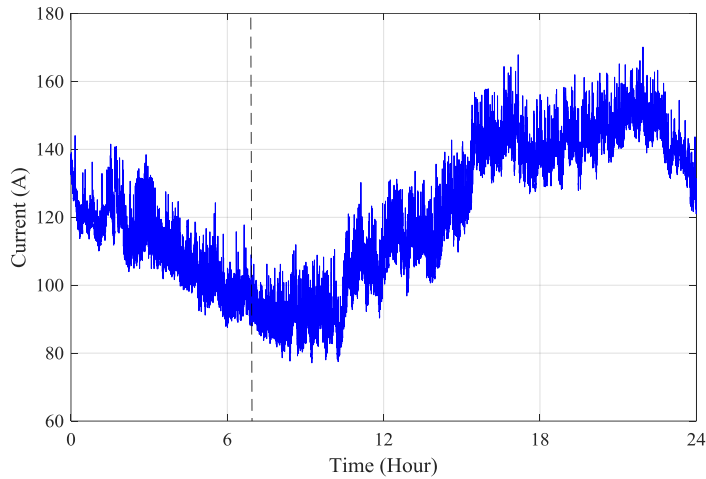
Voltage Angle



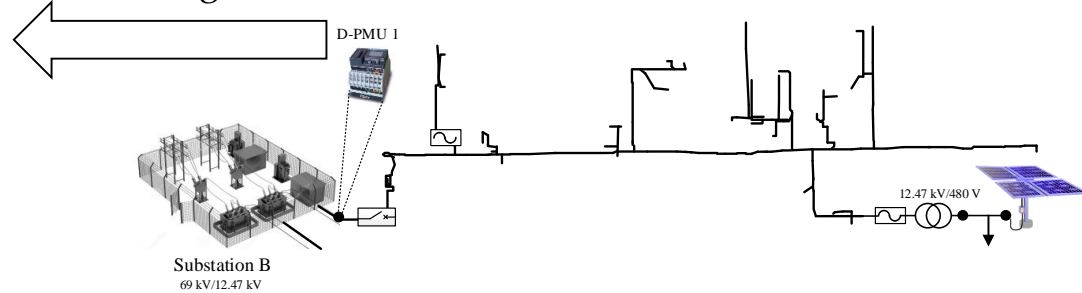
Voltage Magnitude

Events in Distribution Systems

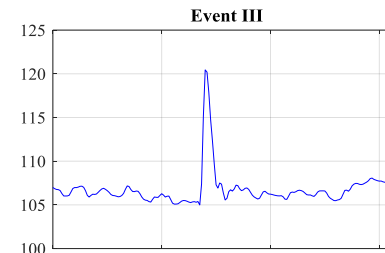
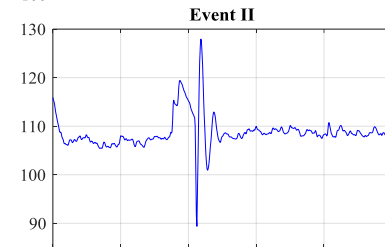
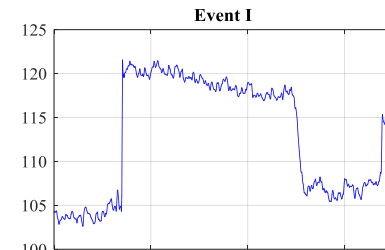
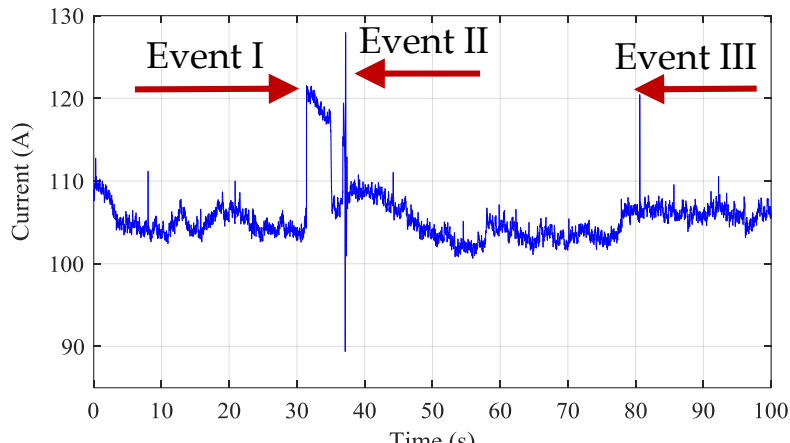
Data points: 10,368,000



Current Magnitude of Phase A



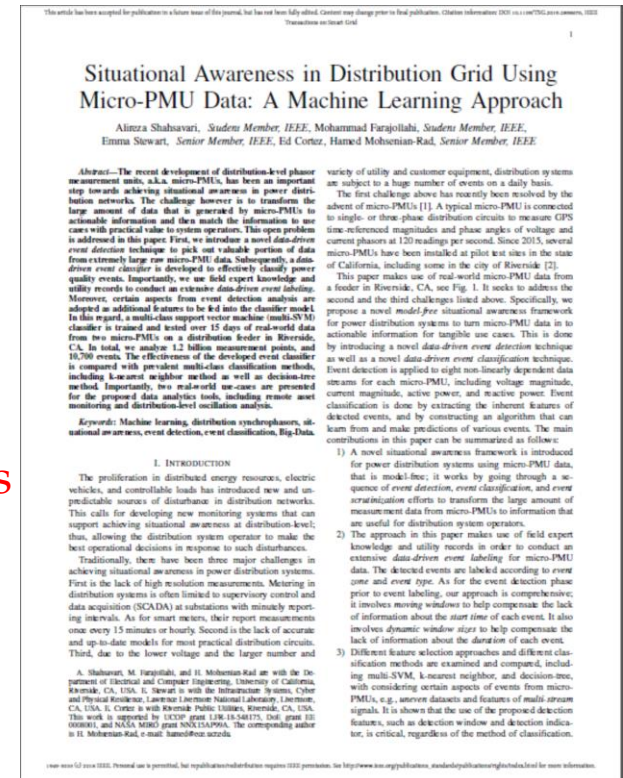
Data Sequence: 100 seconds



Application of Micro-PMUs:

- Capacitor Back Switching
- Fault Analysis
- Lightning Analysis
- Inverter Misoperation
- Event Detection
- Event Classification
- Impedance Calculation
- Topology Identification
- Event Source Location Identification
- ...

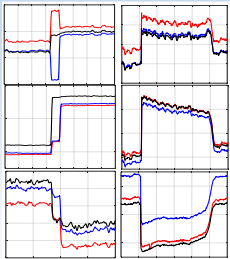
Our Focus:
Situational Awareness
with Application to
Cybersecurity



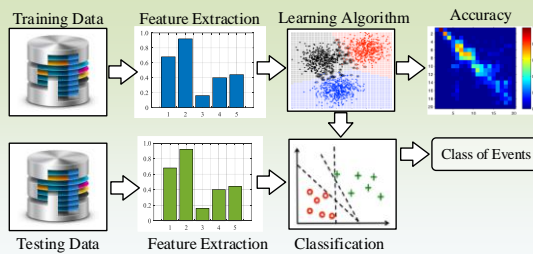
IEEE Power and Energy
IEEE Trans. on Smart Grid, 2019
Magazine, May 2018

Situational Awareness and Cybersecurity in Distribution Systems

Event Detection



Event Classification



Situational Awareness

Perception of Current Situation

Comprehension of current Situation

Projection of Future Status

Cyber Security Analysis in Distribution System

Agenda

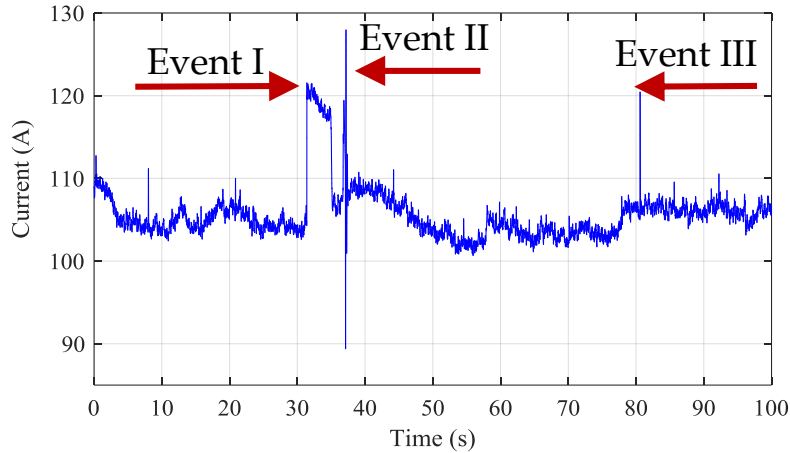
- **Situational Awareness Using Distribution Synchrophasors**
 - Events in Distribution Systems
 - Event Detection
 - Event Classification

- **Practical Use-Cases:**
 - Asset Monitoring
 - Protection System Diagnosis

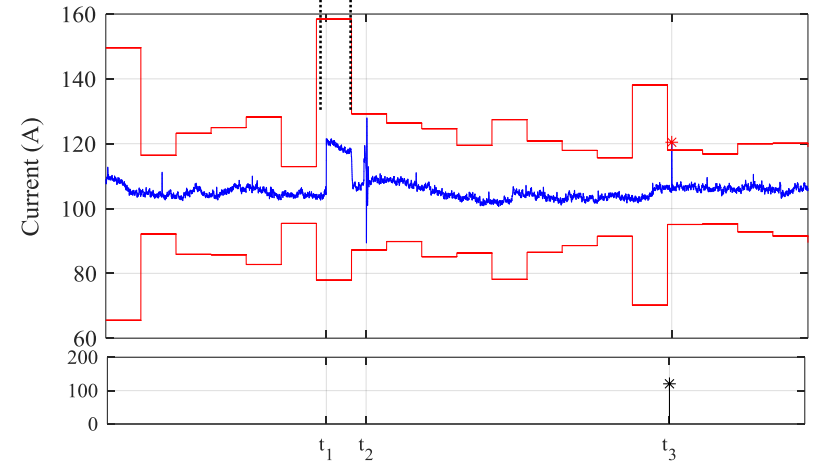
- **Cyber Security & Situational Awareness**

Events Detection: Method I, Absolute Deviation Around Median

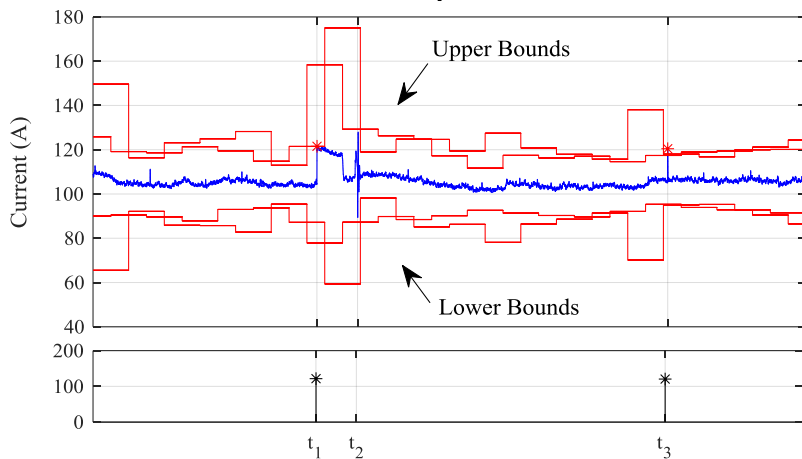
Data Sequence: 100 seconds



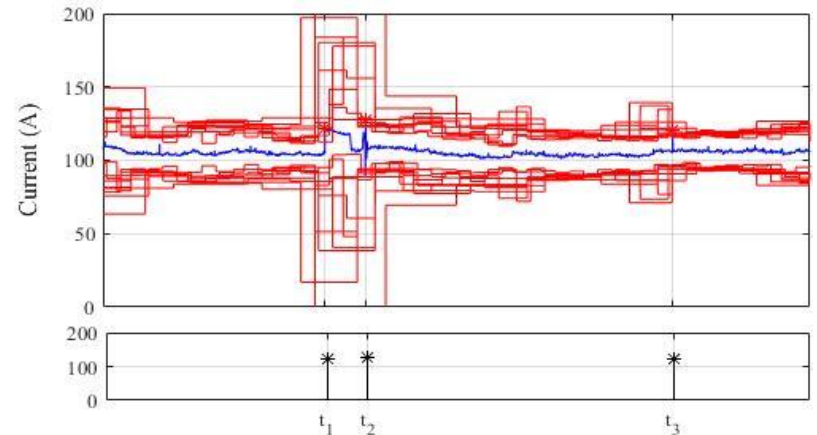
Static Window → ← Window size= 5 seconds



Moving Window

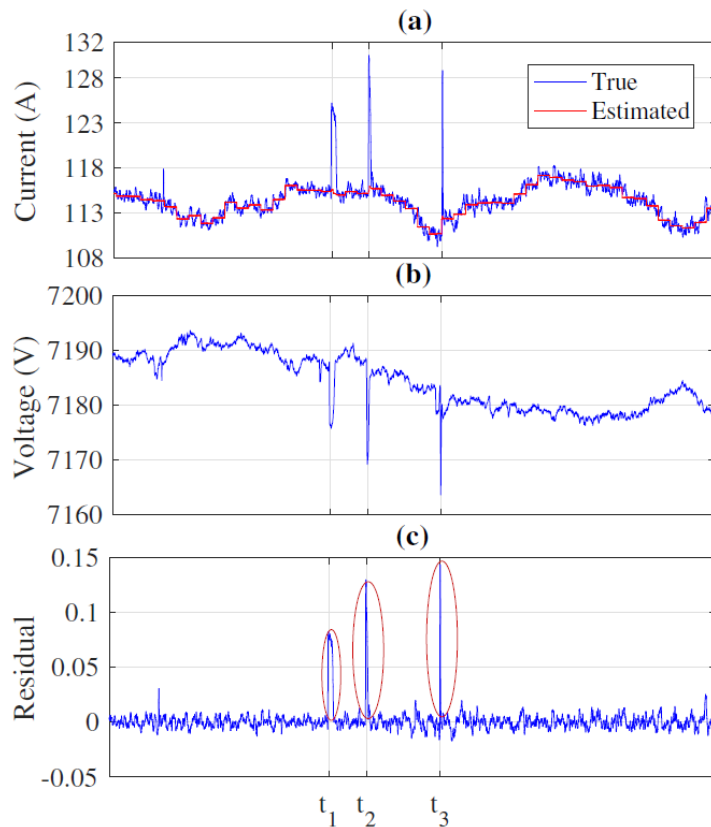


Dynamic Window Size and Moving Window

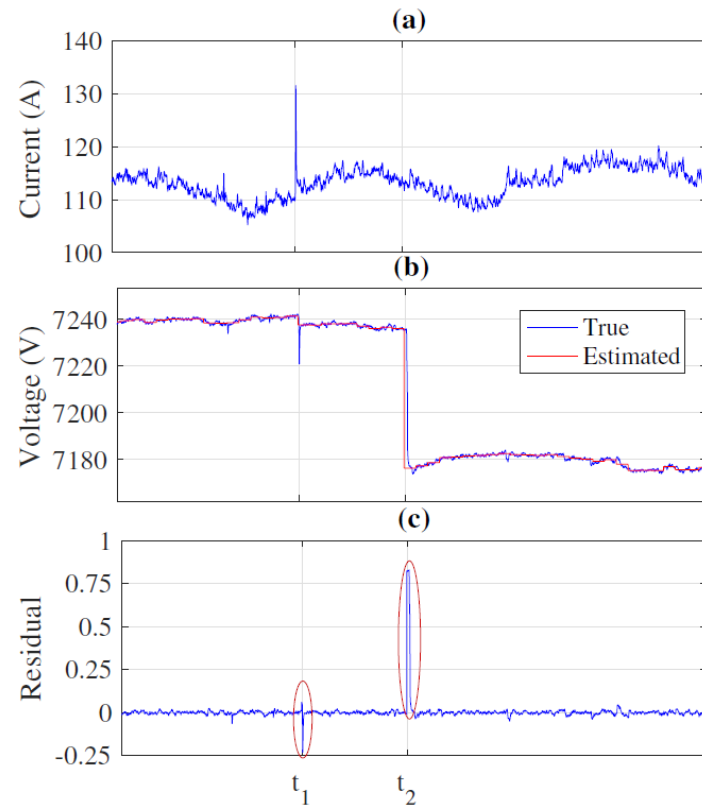


Events Detection: Method II, Residual Test on Non-Linear Estimation

Current Magnitude Estimation



Voltage Magnitude Estimation



Agenda

● **Situational Awareness Using Distribution Synchrophasors**

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● **Use-Cases:**

- Asset Monitoring
- Protection System Diagnosis

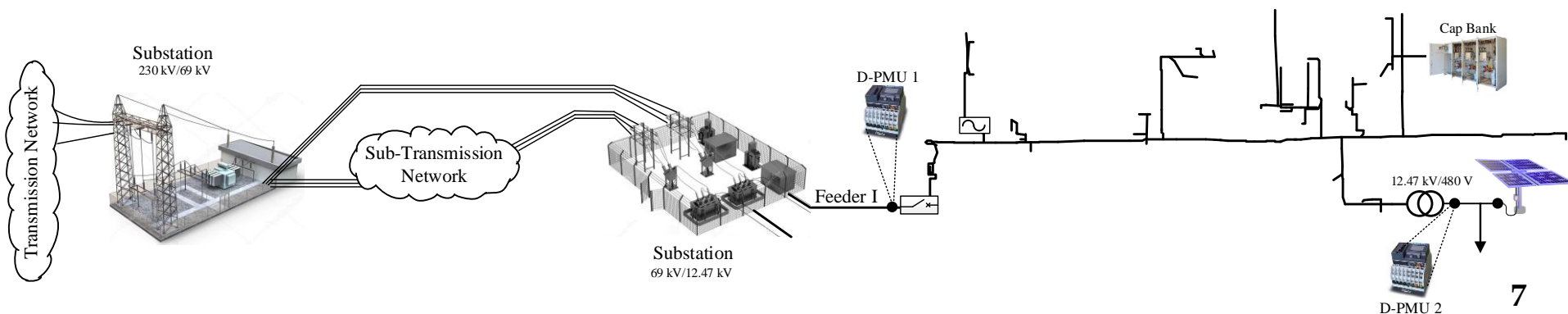
● **Cyber Security & Situational Awareness**

Event Classification

Event Labeling: Field Expert Knowledge and Utility Records

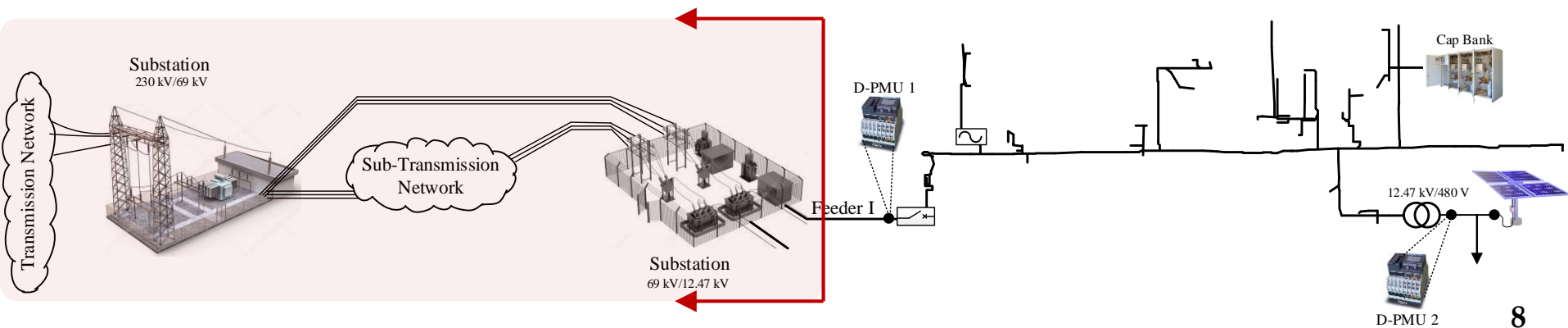
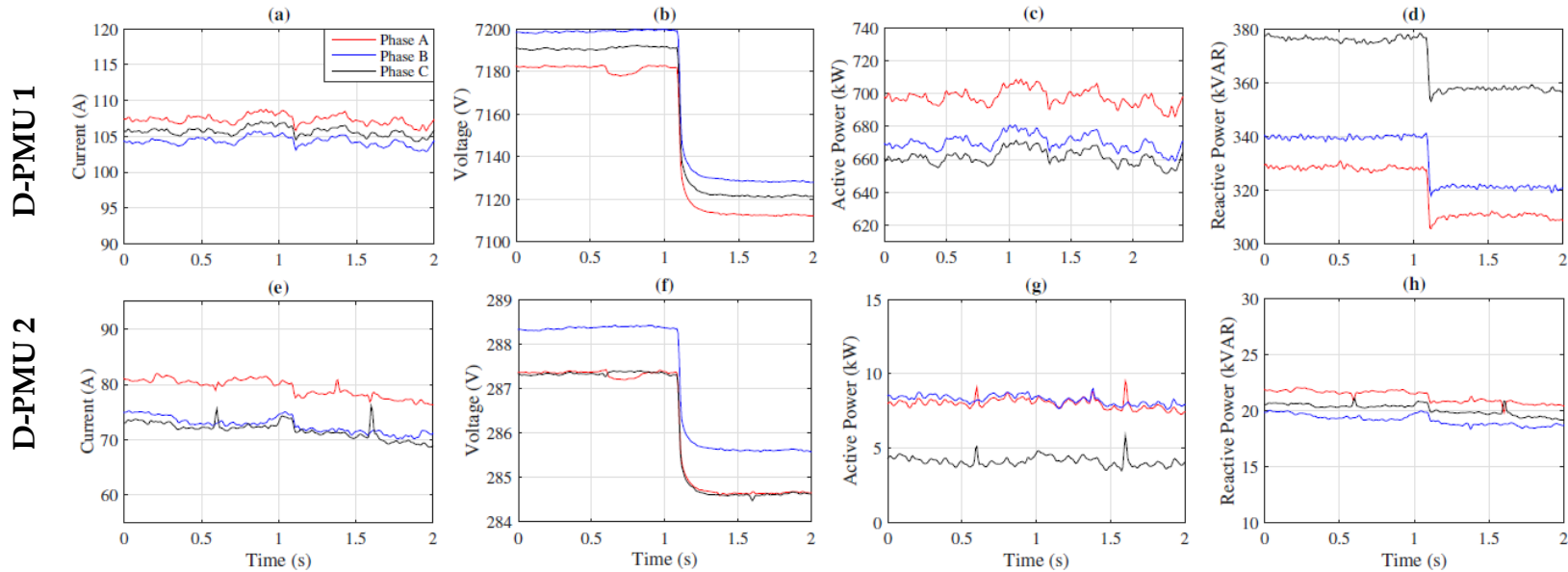
Provisions:

- Field knowledge of utility crew members
- Reviewing more than 1000 utility event logs for one year
- Computer simulation of the understudy feeders.
- Analysis of field data over three years.



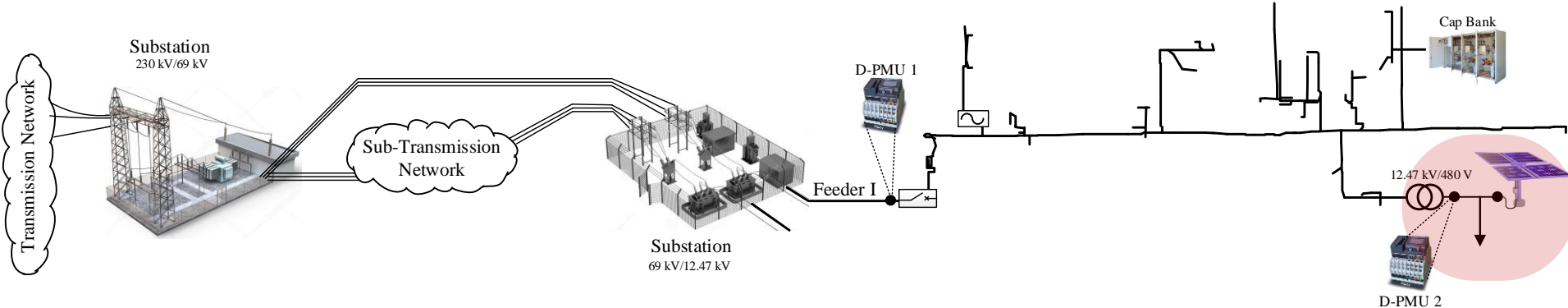
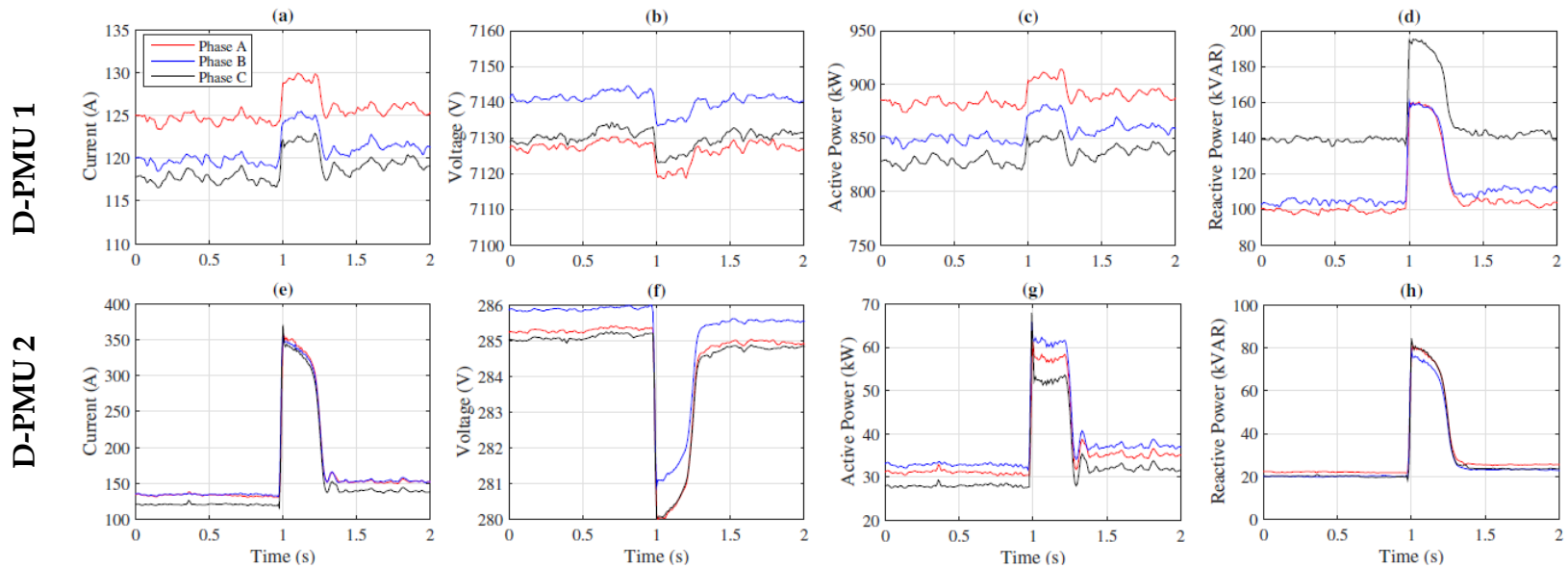
Event Labeling

Class I: Events initiated from upstream of D-PMU 1:



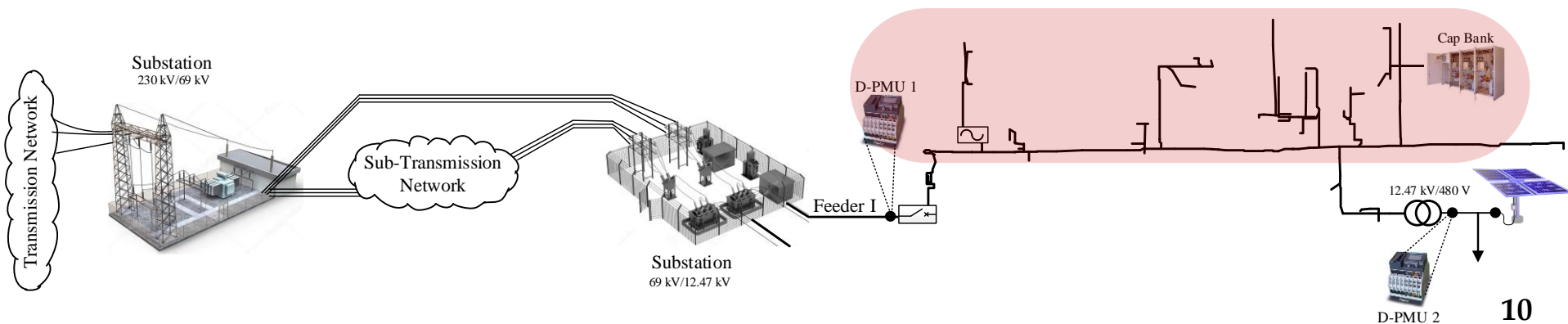
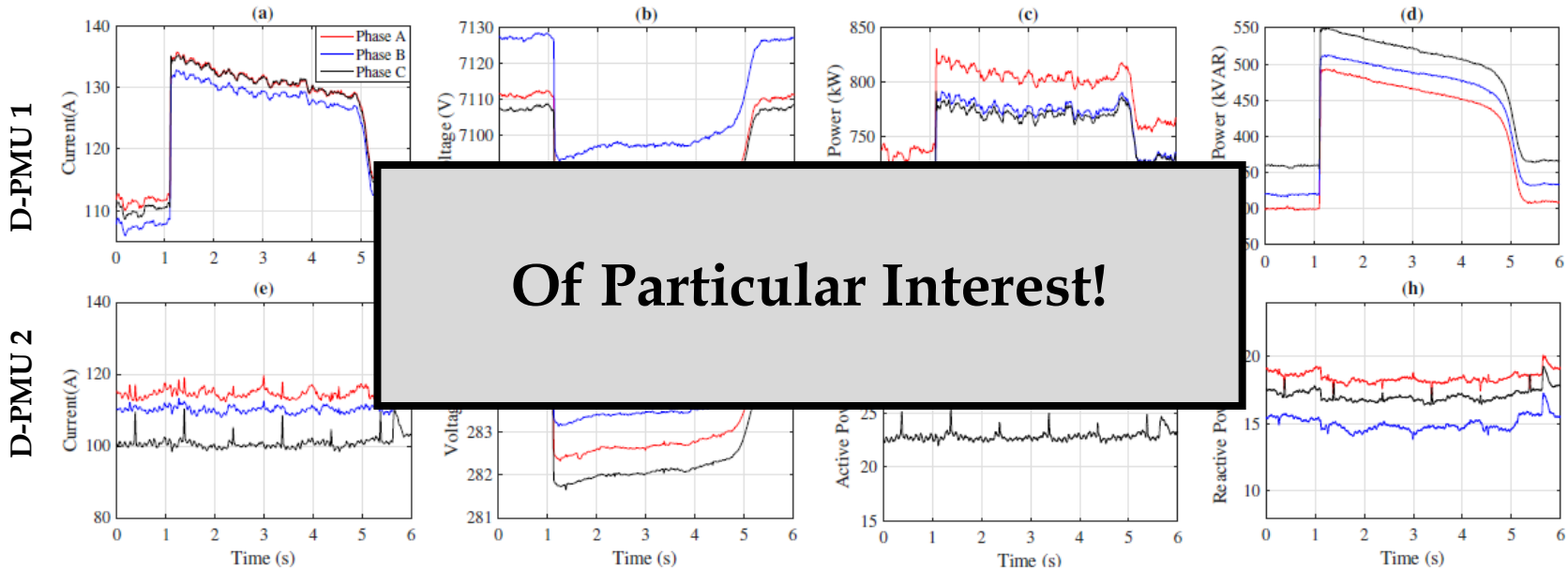
Event Labeling

Class II: Events initiated from downstream of D-PMU 2:



Event Labeling

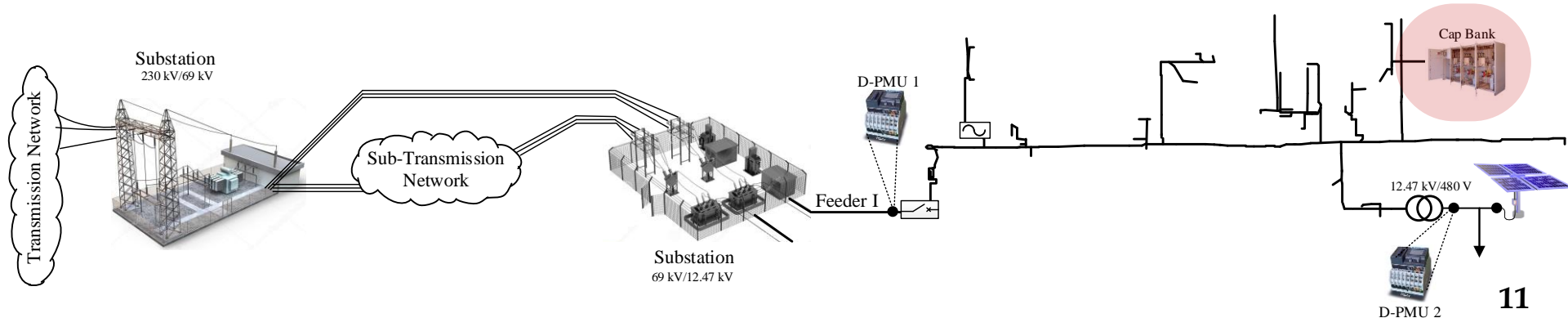
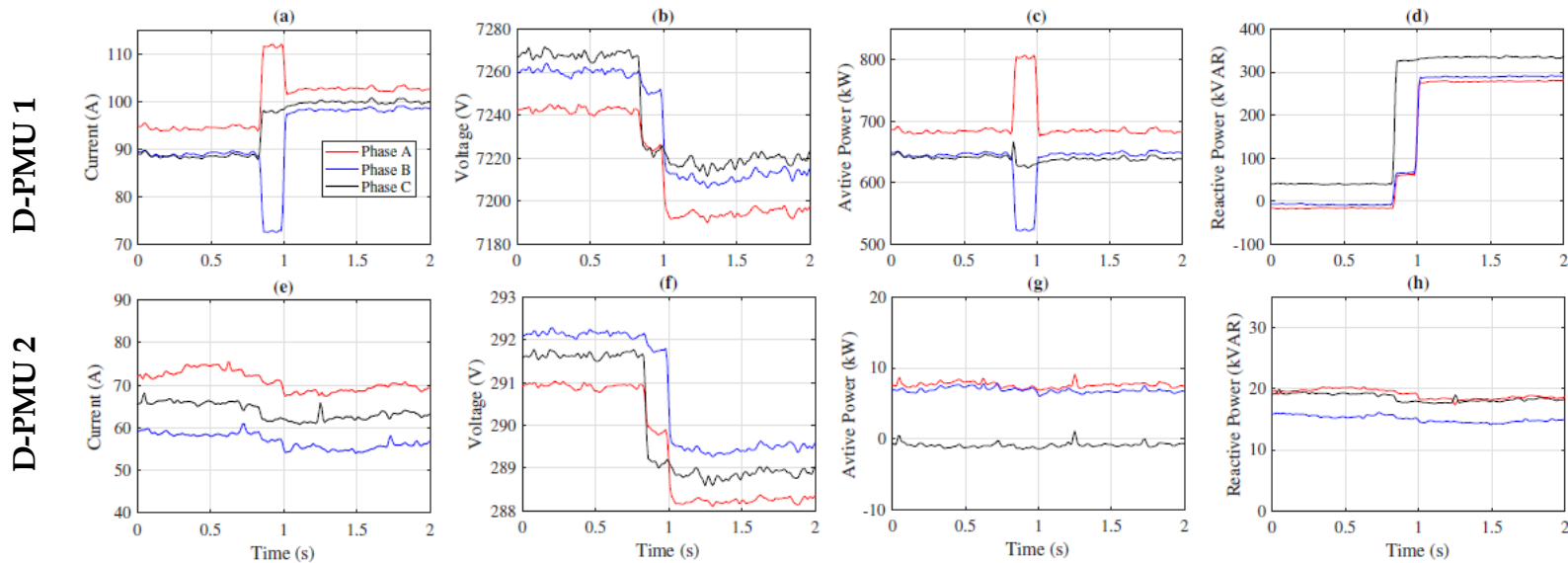
Class III: Events initiated from somewhere between the two D-PMUs:



Event Labeling

Class III: Events initiated from somewhere between the two D-PMUs:

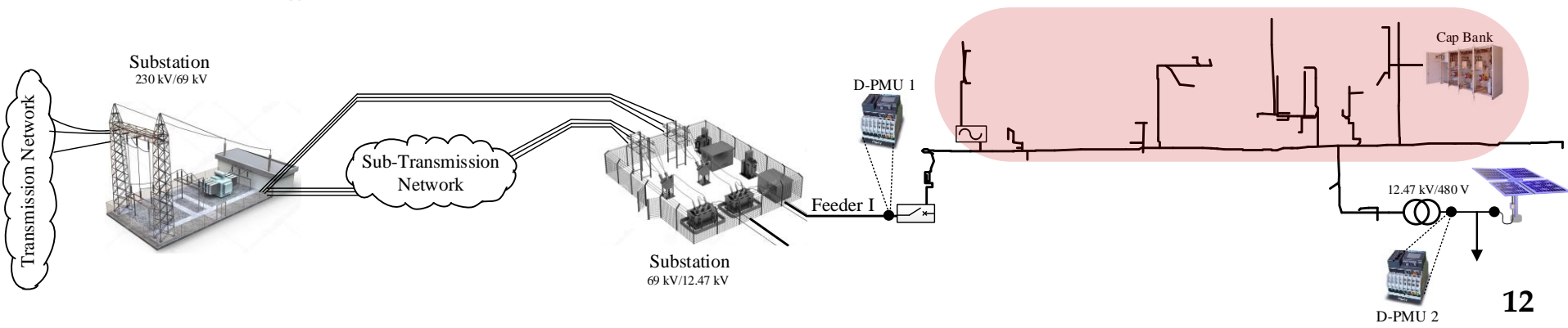
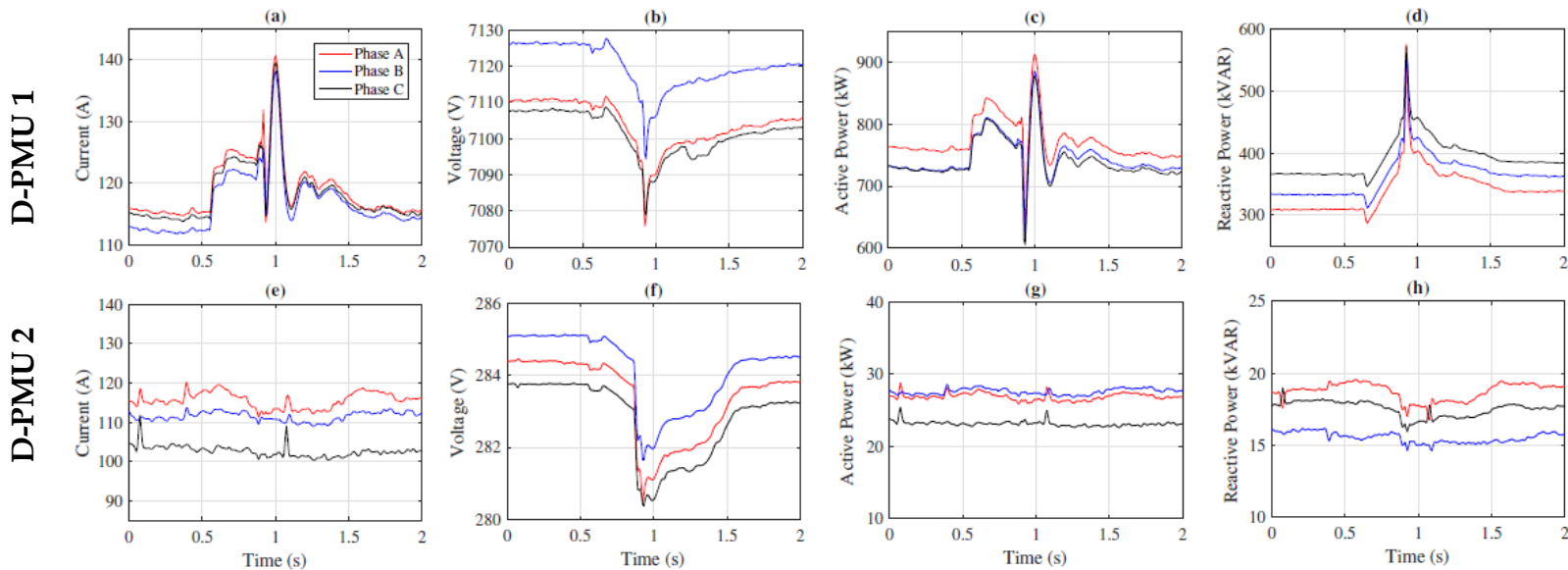
Class III.A: Capacitor Bank Switching



Event Labeling

Class III: Events initiated from somewhere between the two D-PMUs:

Class III.B: Momentary Oscillation



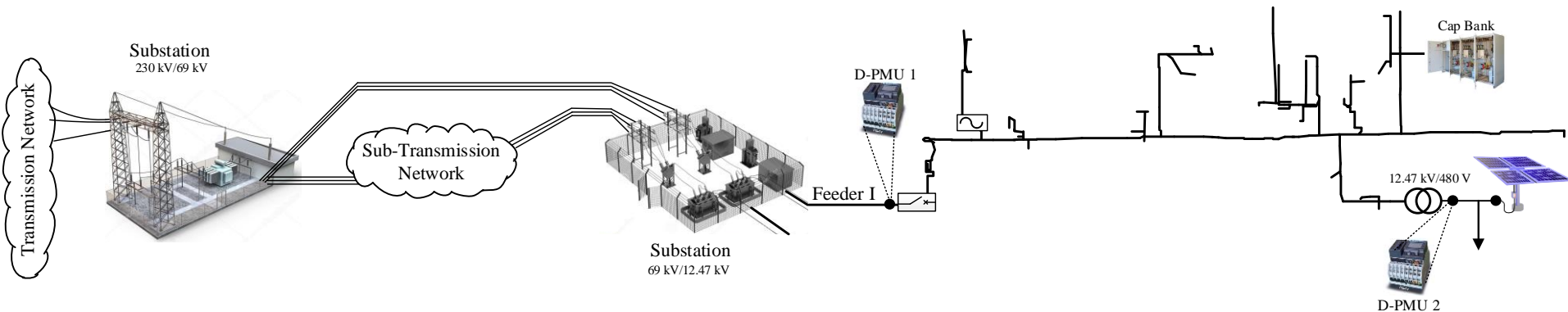
Event Labeling Summary

**Layer I:
Event Zone**

- Class I:** Events initiated from upstream of D-PMU 1
- Class II:** Events initiated from downstream of D-PMU 2
- Class III:** Events initiated from somewhere between the two D-PMUs

**Layer II:
Event Type**

- **Class III.A:** Capacitor Bank Switching Events
- **Class III.B:** Momentary Oscillation
- **Class III.C:** Other Events



Feature Selection

- Single-Stream Features
 - Statistics, e.g., standard deviation
 - Difference, e.g., post-event and pre-event
- Multi-Stream Features
 - Correlation between any two signals
- Event Detection Features
 - Detection window size
 - Detection indicator

We consider data from two D-PMUs: Total number of signals is 8

Feature	Feature Description		Number
Single-stream	Statistics	$\text{std}(D_i)$	8
	Difference	$ d_n - d_1 $	8
Multi-stream	Correlation	$\text{corr}(D_i, D_j)$	28
Detection	Detection Window	ω	1
	Detection Indicator	$\mathbb{I}\{D_i\}$	8

Total number of features: 53

Event Classification Method:

1- Multi-Support Vector Machine (Multi-SVM)

- Binary Decomposition: One-against-all (OvA)

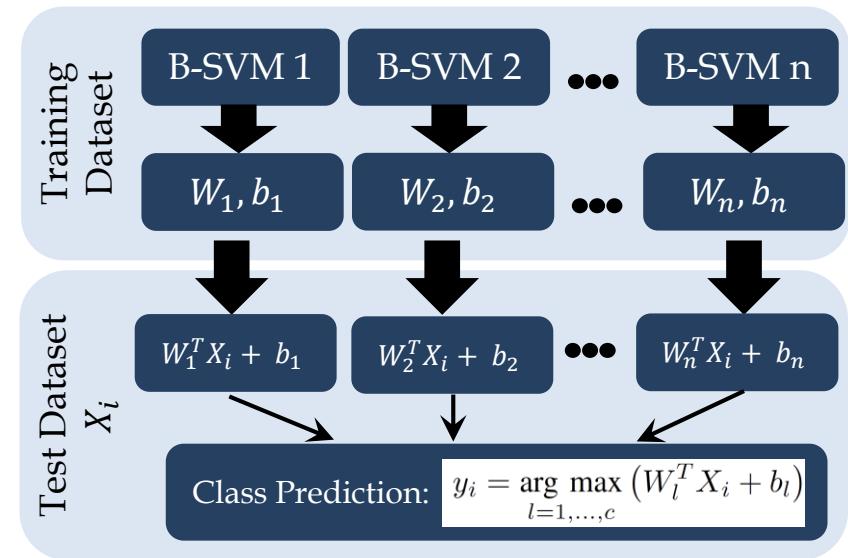
2- k-Nearest Neighbors (k-NN)

3- Decision-Tree (DT)

Summary of Analyzed Database:

- Data from two D-PMUs, during 15 days
- In total, 1.2 billion measurement points
- In total, 10,700 events (only 1% of the data points)
- Number of Events from Event Labeling

- Training dataset almost 4% for both layers



Class I	Class II	Class II
1,802	2,228	6,670

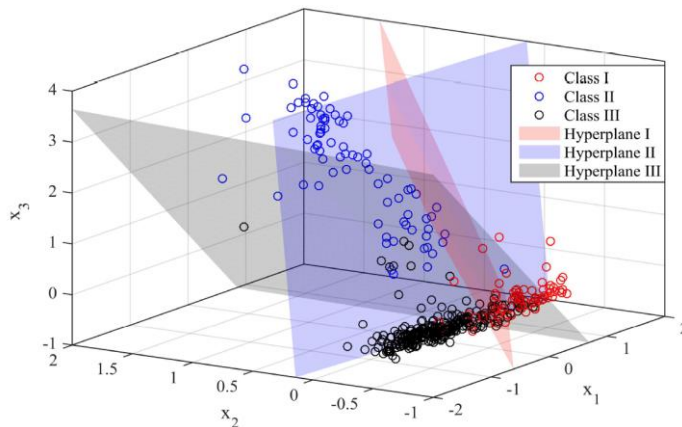


Class III.A	Class III.B	Class III.C
27	43	6,600

Illustrative Example of Multi-SVM Classification Results for Layer I:

Three dominant features among 53 features are selected to train and test classifier.

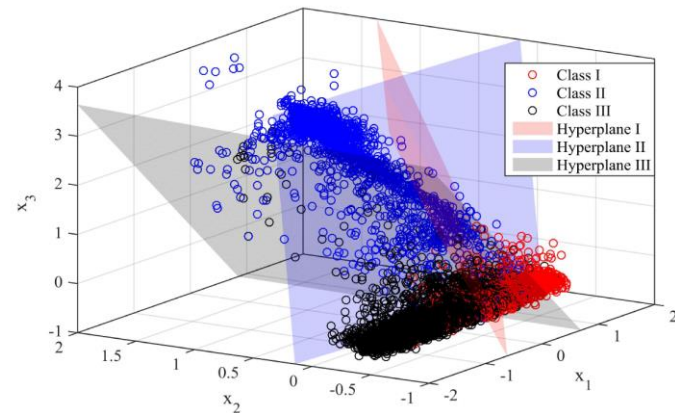
Training



Accuracy: 91.55%

	Class I	Class II	Class III
Class I	82.7% 67	5.8% 5	5.9% 16
Class II	0.0% 0	93.0% 80	0.4% 1
Class III	17.3% 14	1.2% 1	93.7% 254
	Class I	Class II	Class III
	Target Class		

Test



Accuracy: 89.57%

	Class I	Class II	Class III
Class I	70.7% 1216	4.8% 102	6.3% 404
Class II	0.1% 2	94.2% 2018	0.6% 37
Class III	29.2% 503	1.0% 22	93.1% 5958
	Class I	Class II	Class III
	Target Class		

Classification Results for Test data of Layer I:

Impact of Detection Features

Multi-SVM Classifier

k-NN Classifier

Decision Tree Classifier

Without Detection Features

With Detection Features

Accuracy: 98.16%

Predicted Class	Class I	Class II	Class III
Class I	95.2% 1639	2.0% 43	0.2% 14
Class II	0.7% 12	97.5% 2089	0.6% 40
Class III	4.1% 70	0.5% 10	99.2% 6345
	Class I	Class II	Class III
	Target Class		

Accuracy: 97.82%

Predicted Class	Class I	Class II	Class III
Class I	97.3% 1675	2.7% 58	0.8% 54
Class II	0.2% 3	96.3% 2063	0.7% 45
Class III	2.5% 43	1.0% 21	98.5% 6300
	Class I	Class II	Class III
	Target Class		

Accuracy: 92.74%

Predicted Class	Class I	Class II	Class III
Class I	74.6% 1284	0.3% 6	0.6% 41
Class II	0.5% 8	88.7% 1899	0.4% 24
Class III	24.9% 429	11.1% 237	99.0% 6334
	Class I	Class II	Class III
	Target Class		

Accuracy: 99.89%

Predicted Class	Class I	Class II	Class III
Class I	100.0% 1721	0.2% 5	0.0% 3
Class II	0.0% 0	99.7% 2136	0.0% 2
Class III	0.0% 0	0.0% 1	99.9% 6394
	Class I	Class II	Class III
	Target Class		

Accuracy: 99.20%

Predicted Class	Class I	Class II	Class III
Class I	99.2% 1708	1.1% 24	0.3% 19
Class II	0.2% 3	98.6% 2111	0.3% 19
Class III	0.6% 10	0.3% 7	99.4% 6361
	Class I	Class II	Class III
	Target Class		

Accuracy: 99.48%

Predicted Class	Class I	Class II	Class III
Class I	100.0% 1721	0.0% 1	0.0% 1
Class II	0.0% 0	97.6% 2091	0.0% 1
Class III	0.0% 0	2.3% 50	100.0% 6397
	Class I	Class II	Class III
	Target Class		

Classification Results for Test data of Layer II

Multi-SVM Classifier

Accuracy: 99.75%

Predicted Class	Class A	100.0% 25	0.0% 0	0.1% 7
	Class B	0.0% 0	94.9% 37	0.1% 7
	Class C	0.0% 0	5.1% 2	99.8% 6321
	Target Class	Class A	Class B	Class C

k-NN Classifier

Accuracy: 99.87%

Predicted Class	Class A	100.0% 25	2.6% 1	0.0% 2
	Class B	0.0% 0	97.4% 38	0.1% 5
	Class C	0.0% 0	0.0% 0	99.9% 6328
	Target Class	Class A	Class B	Class C

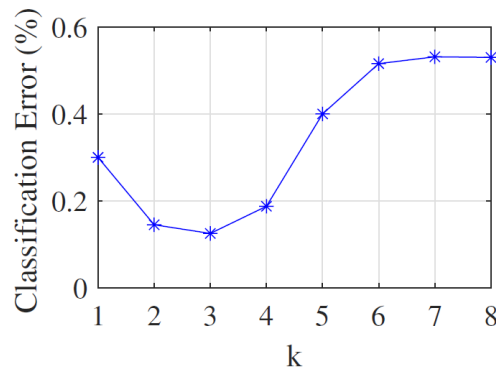
Decision Tree Classifier

Accuracy: 99.23%

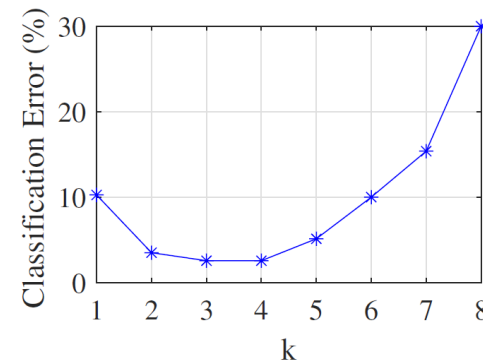
Predicted Class	Class A	0.0% 0	0.0% 0	0.0% 0
	Class B	88.0% 22	100.0% 39	0.4% 24
	Class C	12.0% 3	0.0% 0	99.6% 6311
	Target Class	Class A	Class B	Class C

- The performance of k-NN classifier is highly sensitive to the choice of parameter k.

Overall Classification Error (%)



Classification Error in Class III.B (%)



Agenda

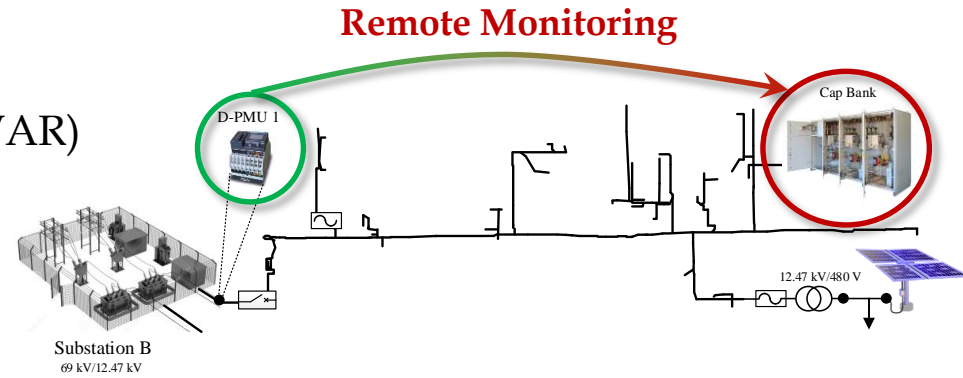
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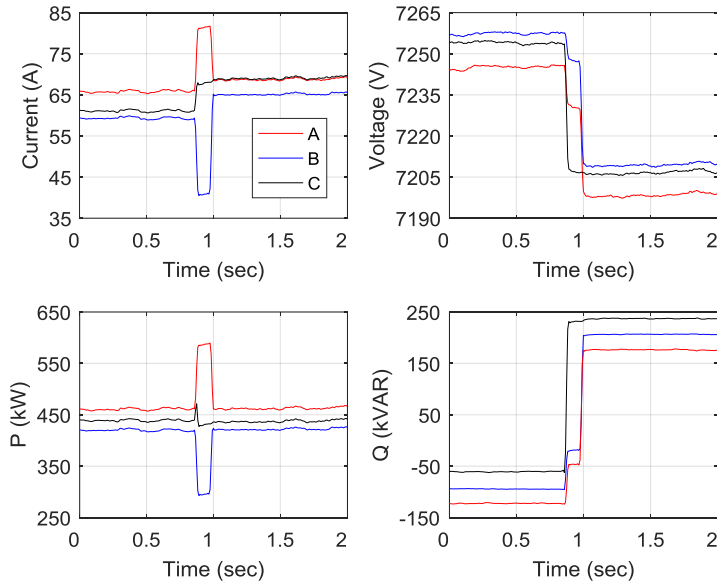
- **Cyber Security & Situational Awareness**

Three-Phase Switched Capacitor Bank:

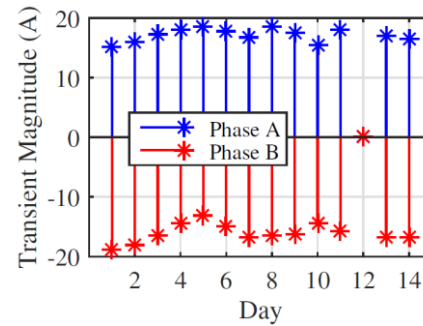
- Switched capacitor bank three-phase (900 kVAR)
 - Controller: Volt-VAR Controller.



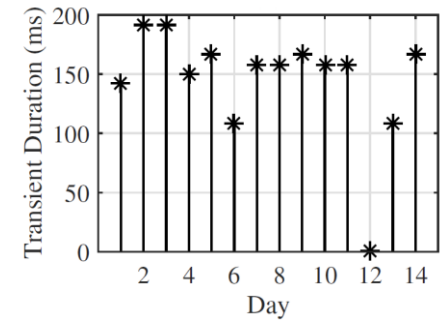
Switch off Event



Transient Current Magnitude

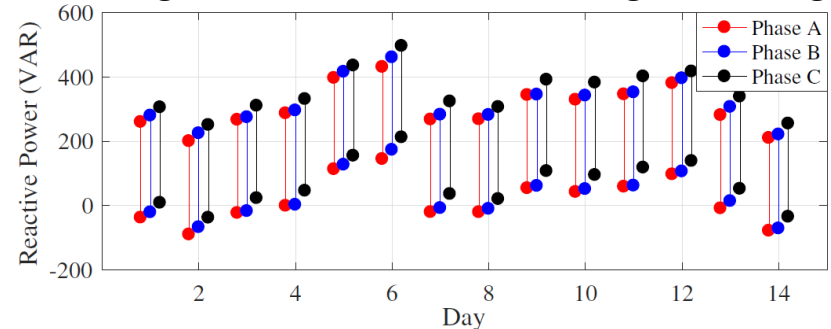


Duration of Transition



- **Two-Step 3-Phase Switch**
 - Step 1: Phase C (Zero Crossing)
 - Step 2: Phase A/B (Possible Malfunction)
- **Slightly Unbalanced Operation**
 - Likely fuse blowing on A and C

Change in Reactive Power During Switching



Agenda

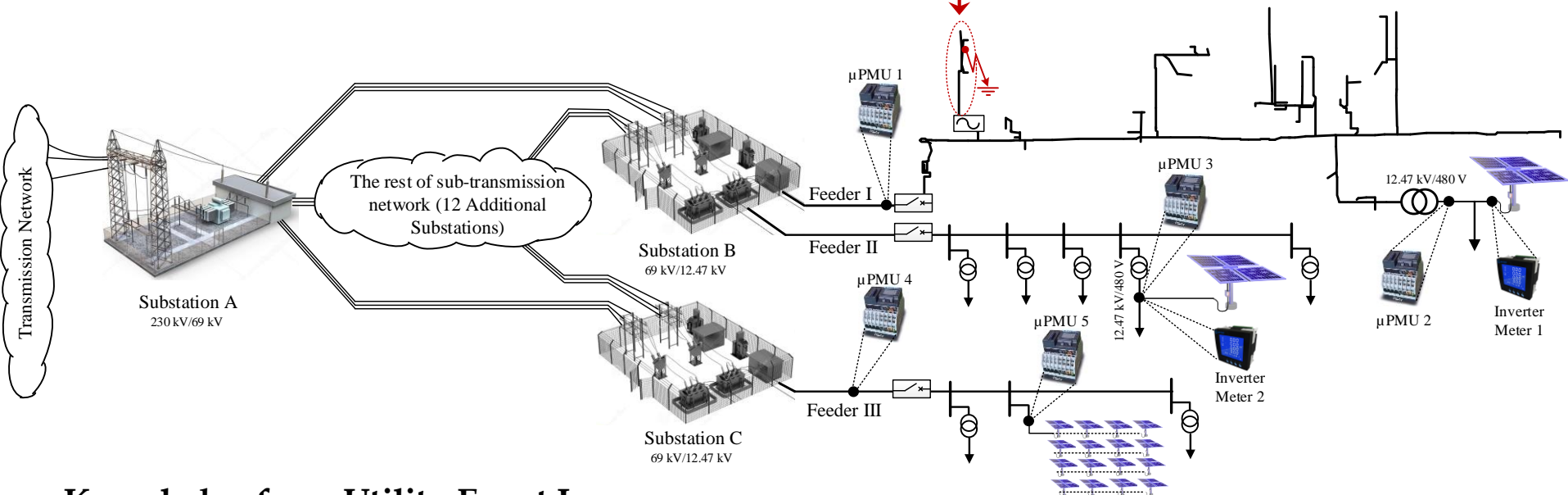
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Protection System Diagnosis

Momentary Fault Occurrence:



Knowledge from Utility Event Log:

- **Fault Location:** Feeder I, in November 2016, almost 0.3 mile away from Substation B.
- **Event Source:** Blown fuse is founded in phase B and a diseased bird on pole.

We are interested in answering the following questions

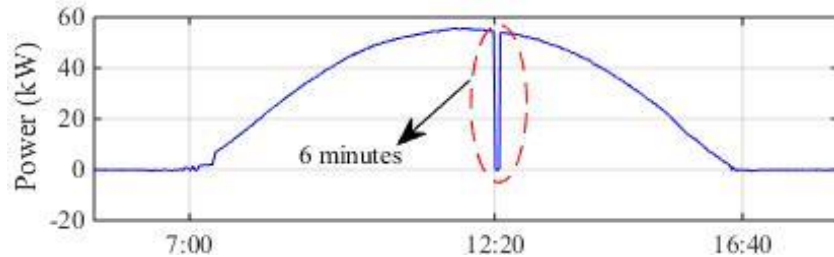
- What is the time-line of the fault?
- Possible miscoordination between lateral fuse and auto-recloser?
- Possible miscoordination between auto-recloser anti-islanding relay?

Fault Time-Line:

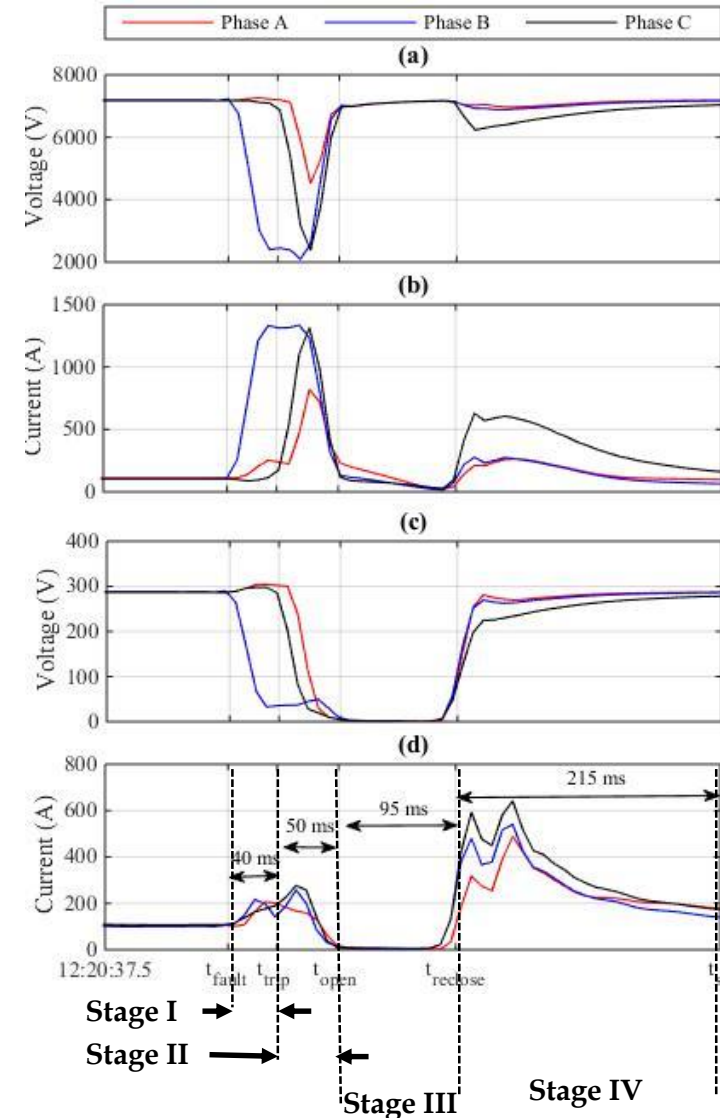
Stage I	<ul style="list-style-type: none"> Voltage on phase B drops Deviation in voltage on phases A and C
Stage II	<ul style="list-style-type: none"> Recloser sends trip command to the circuit breaker Three-phase circuit breaker operation transient
Stage III	<ul style="list-style-type: none"> Feeder I is isolated Current of D-PMU 1 decreases to zero (RLC circuit)
Stage IV	<ul style="list-style-type: none"> First shot of recloser is finished Feeder is reconnected to substation



Recloser-fuse coordination is Verified



Recloser Anti-islanding protection Miscoordination



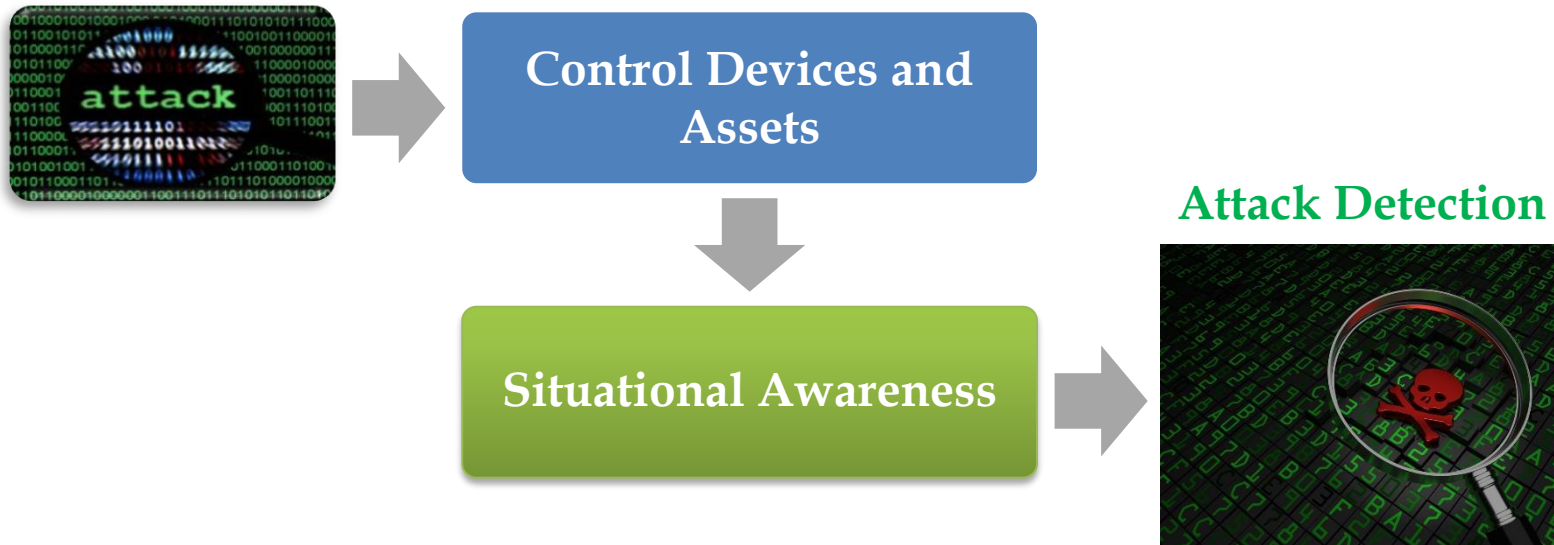
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- **Cyber Security & Situational Awareness**

- Aspect 1: Attack Detection through Situational Awareness



- Aspect 2: Vulnerability in Sensors and Monitoring System



Mohasinia's Research

References:

- [1] A. Shahsavari, M. Farajollahi, E. Stewart, E. Cortez, H. Mohsenian Rad, "Situational Awareness in Distribution Grid Using Micro-PMU Data: A Machine Learning Approach", *IEEE Trans. on Smart Grid*, 2019.
- [2] M. Farajollahi, A. Shahsavari, E. Stewart, H. Mohsenian Rad, "Locating the Source of Events in Power Distribution Systems Using Micro-PMU Data", *IEEE Trans. on Power Systems*, vol.33, pp. 6343 - 6354, May. 2018.
- [3] A. Shahsavari, A. Sadeghi-Mobarakeh, E. Stewart, E. Cortez, L. Alvarez, F. Megala, H. Mohsenian Rad, "Distribution Grid Reliability versus Regulation Market Efficiency: An Analysis based on Micro-PMU Data," *IEEE Trans. on Smart Grid*, vol. 8, pp. 2916 - 2925, Jun. 2017.
- [4] A. Shahsavari, M. Farajollahi, E. Stewart, H. Mohsenian Rad, "A Machine Learning Approach to Event Analysis in Distribution Feeders Using Distribution Synchrophasors," in *Proc. of IEEE Smart Grid Synchronized Measurements and Analytics*, College Station, TX, May. 2019.
- [5] A. Shahsavari, M. Farajollahi, E. Stewart, C. Roberts, F. Megala, L. Alvarez, E. Cortez, H. Mohsenian Rad, "Autopsy on Active Distribution Networks: A Data-Driven Fault Analysis Using Micro-PMU Data," in *Proc. of IEEE PES NAPS*, Morgantown, WV, Sep. 2017.
- [6] A. Shahsavari, M. Farajollahi, E. Stewart, A. Meier, L. Alvarez, E. Cortez, H. Mohsenian Rad, "A Data-Driven Analysis of Capacitor Bank Operation at a Distribution Feeder Using Micro-PMU Data," in *Proc. of IEEE Power & Energy Society Conference on Innovative Smart Grid Technologies (ISGT)*, Washington, DC, Apr. 2017.
- [7] A. Shahsavari, M. Farajollahi, E. Stewart, C. Roberts, H. Mohsenian Rad, "A Data-Driven Analysis of Lightning-Initiated Contingencies at a Distribution Grid with a PV Farm Using Micro-PMU Data," in *Proc. of IEEE PES Society NAPS*, Morgantown, WV, Sep. 2017.
- [8] A. Shahsavari, A. Sadeghi, E. Stewart, H. Mohsenian Rad, "Distribution Grid Reliability Analysis Considering Regulation Down Load Resources via Micro-PMU Data," in *proc. of IEEE International Conference on Smart Grid Communications*, Sydney, Australia, Nov. 2016.

Thank You!

Alireza Shahsavari