

Tutorial: Synchro-waveform Data Analytics and Applications

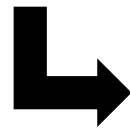
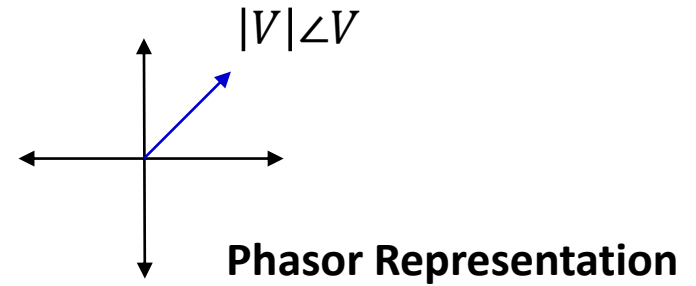
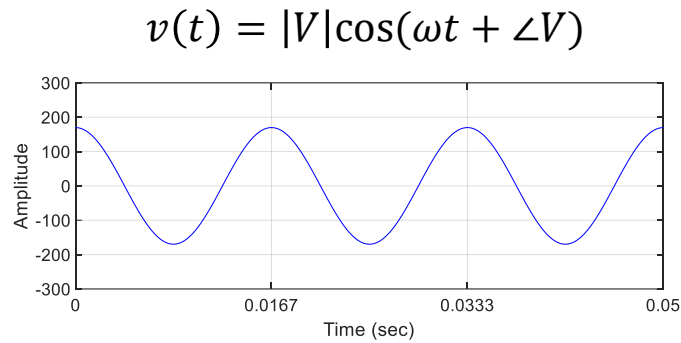
IEEE PES Subcommittee on Big Data Analytics

November 29, 2022

Hamed Mohsenian-Rad, *Ph.D., IEEE Fellow*

Professor and Bourns Family Faculty Fellow
Department of Electrical Engineering, University of California, Riverside
Associate Director, Winston Chung Global Energy Center

Motivation

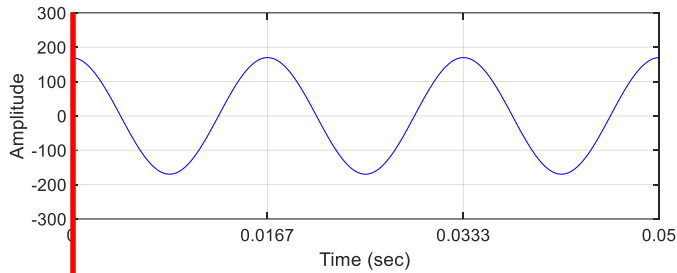


Sensor: Phasor Measurement Unit (PMU)

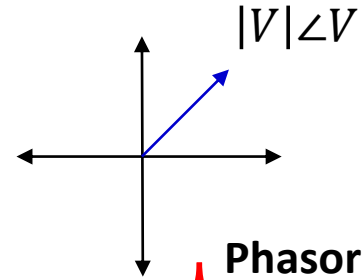
FFT \mapsto Fundamental Component

Motivation

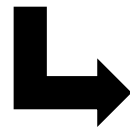
$$v(t) = |V|\cos(\omega t + \angle V)$$



**Time
Reference**



Phasor Representation
Synchro



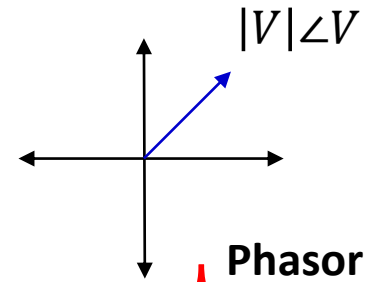
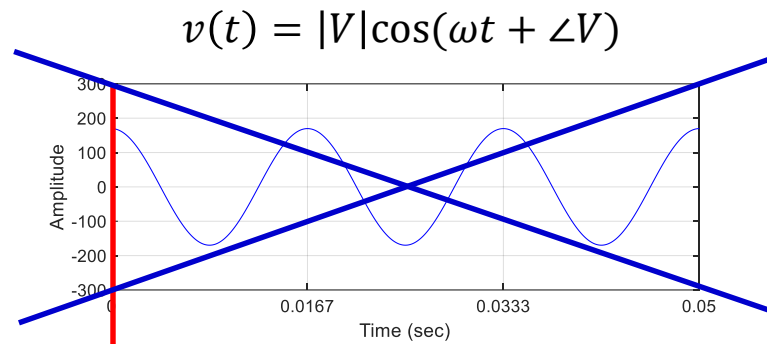
Sensor: Phasor Measurement Unit (PMU)

FFT \mapsto Fundamental Component



GPS Satellites

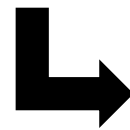
Motivation



Phasor Representation

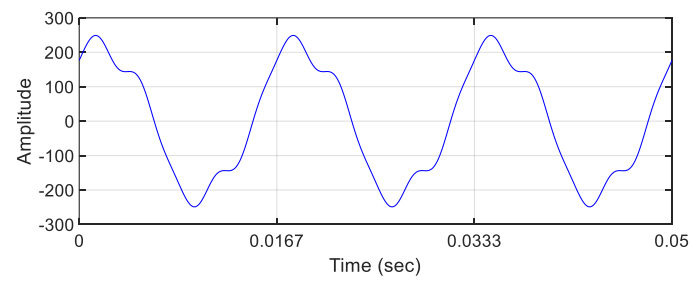
Synchro

Time Reference



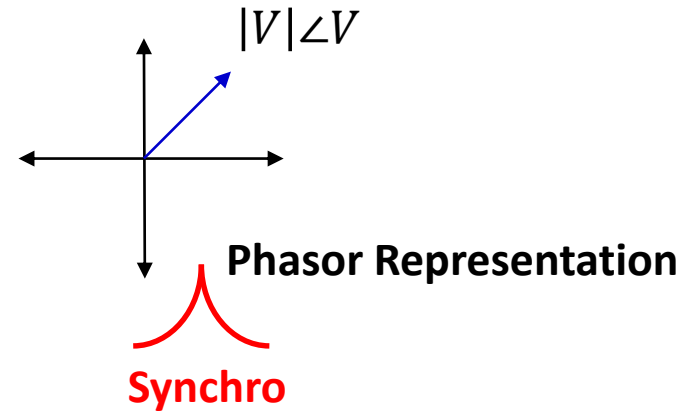
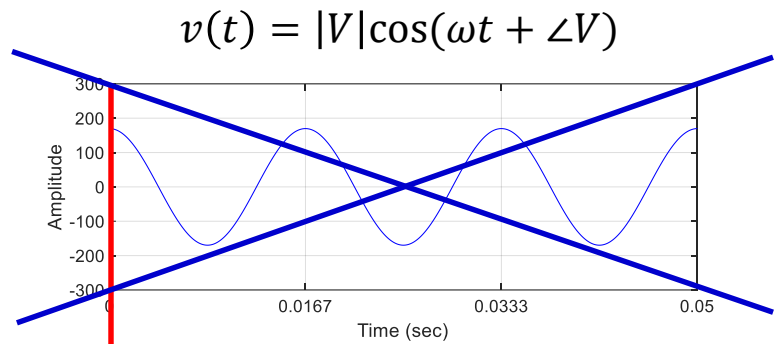
Sensor: Phasor Measurement Unit (PMU)

FFT \mapsto Fundamental Component

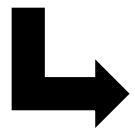


GPS Satellites

Motivation

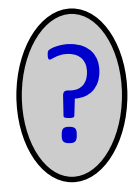
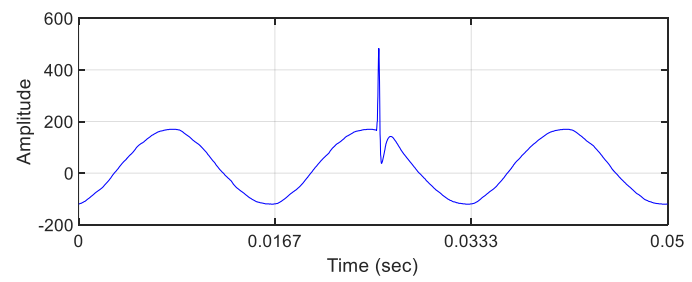


Time Reference



Sensor: Phasor Measurement Unit (PMU)

FFT \mapsto Fundamental Component

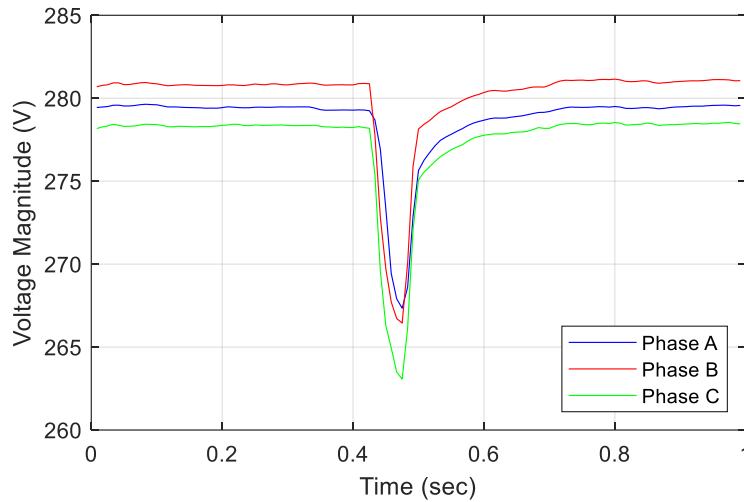


GPS Satellites

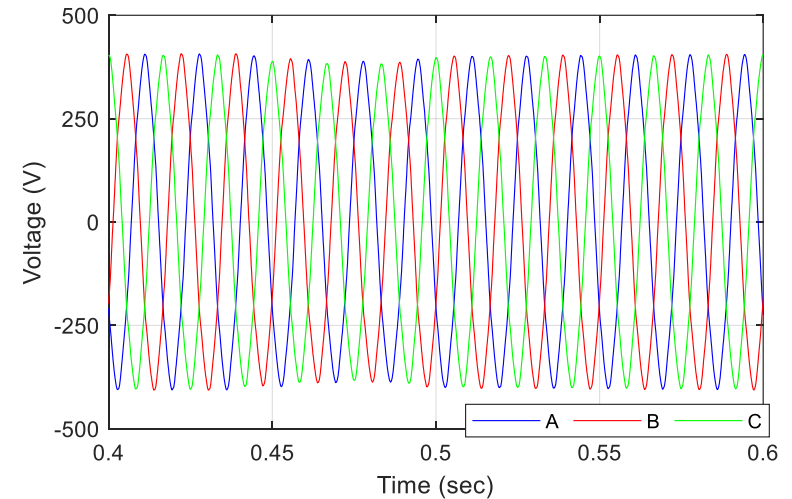
- **Fundamentals**
 - Waveform: Real-World Examples
 - Waveform Measurement Unit
 - Synchro-Waveforms
- **Data-Analytics Methodologies**
 - Detection
 - Location Identification
 - Characterization and Classification
- **Applications**
- **Further Reading**

Waveform: Real-World Examples

- Example 1 (Voltage Sag):



Phasor (Magnitude)

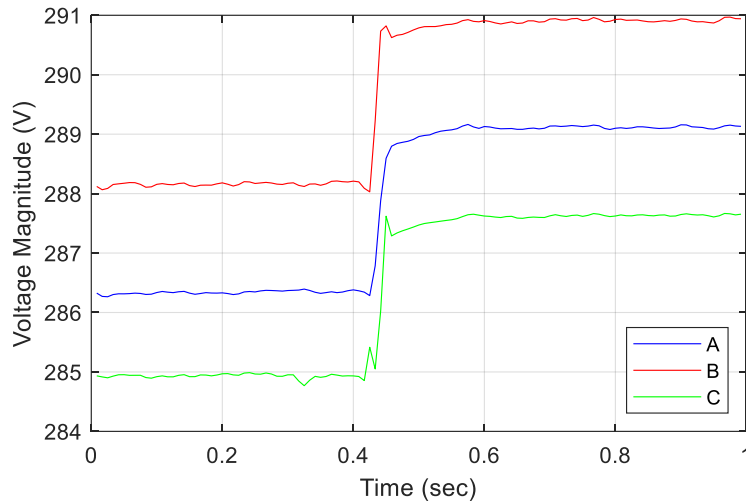


Waveform

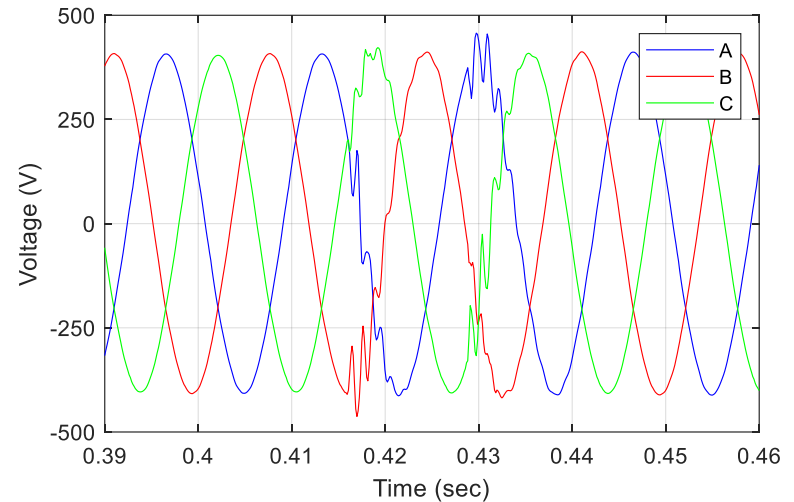
- Looking at voltage waveform is *not necessary* in this example.

Waveform: Real-World Examples

- Example 2 (Resonance):



Phasor (Magnitude)

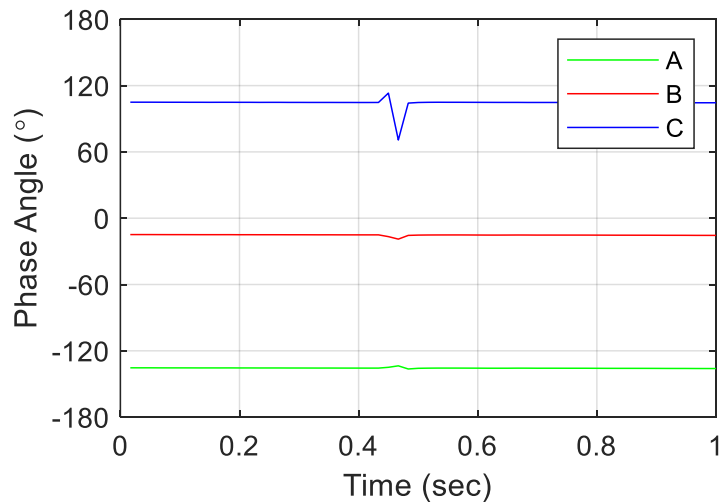
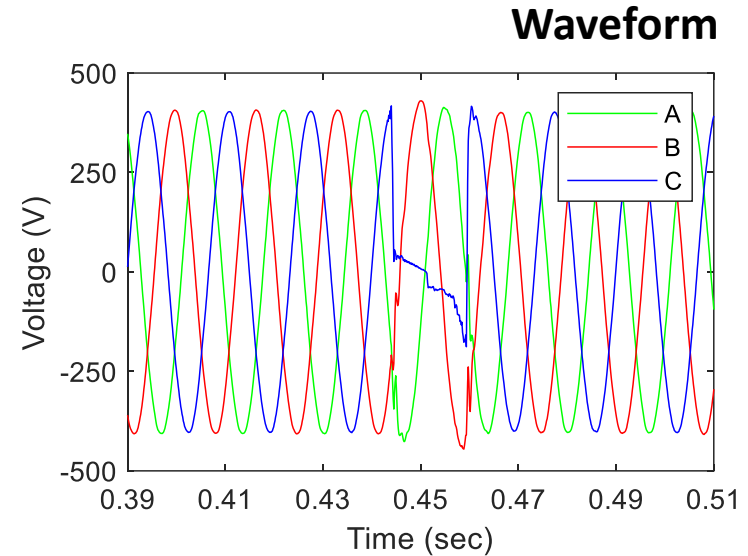
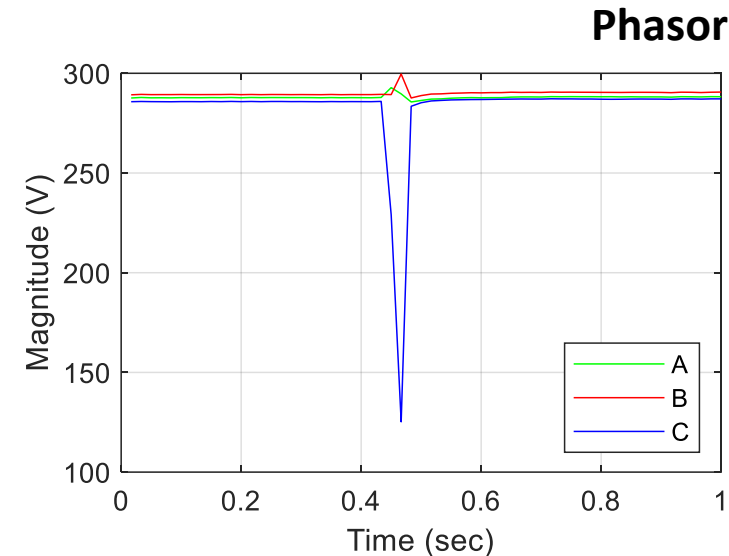


Waveform

- We *cannot see* the high-frequency resonance in the phasors.

Waveform: Real-World Examples

- Example 3 (Fault):



- Waveforms show much more details in this example.

Waveform Measurement Unit

- The device to measure voltage and current waveform:
 - **WMU**: Waveform Measurement Unit¹

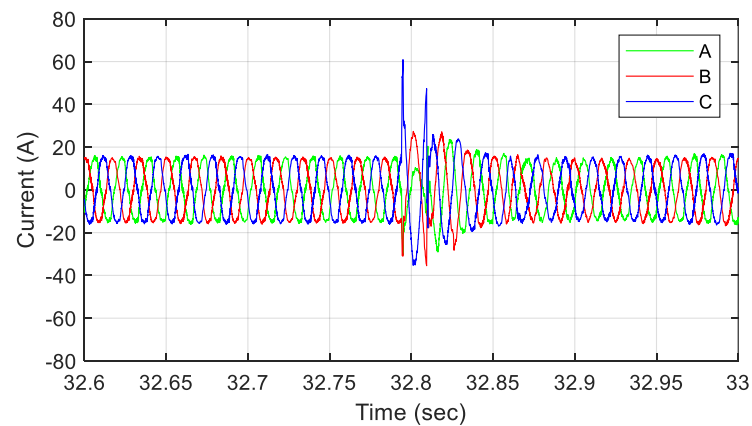
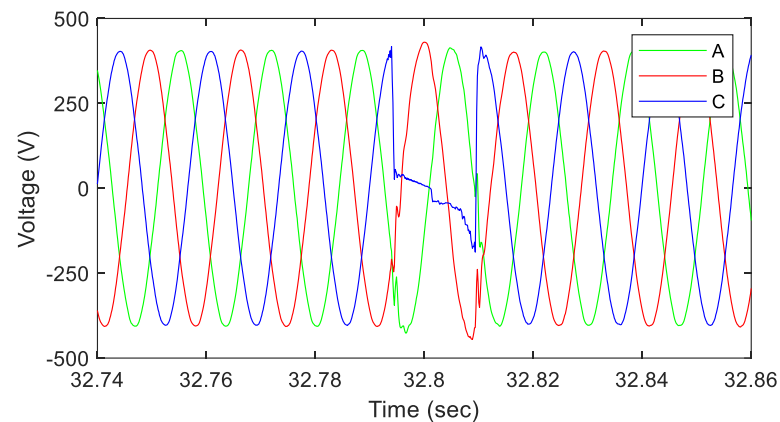
(Compare it with **PMU**: Phasor Measurement Unit)

- WMU is a generic term. The actual sensor might be called:
 - Power Quality Meter
 - Digital Fault Recorder (DFR) (They all measure waveform)
 - Point-on-Wave (POW) Sensor

¹ H. Mohsenian-Rad, *Smart Grid Sensors: Principles and Applications*, Cambridge University Press, April 2022.

Waveform Measurement Unit

- WMUs can measure both **voltage** and **current** waveforms:
- Measured by the same WMUs (over 12 terminals):



Synchro-Waveforms

- Two Concepts:

Synchro-Phasors = Phasors + Time Synchronization

Synchro-Waveforms = Waveforms + Time Synchronization

}
GPS Clock

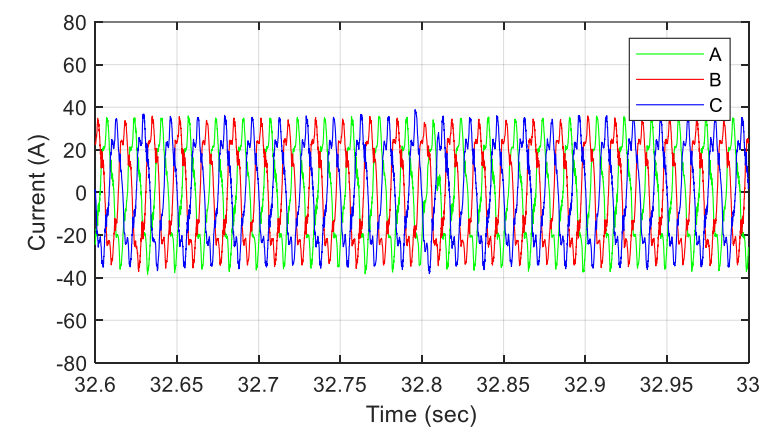
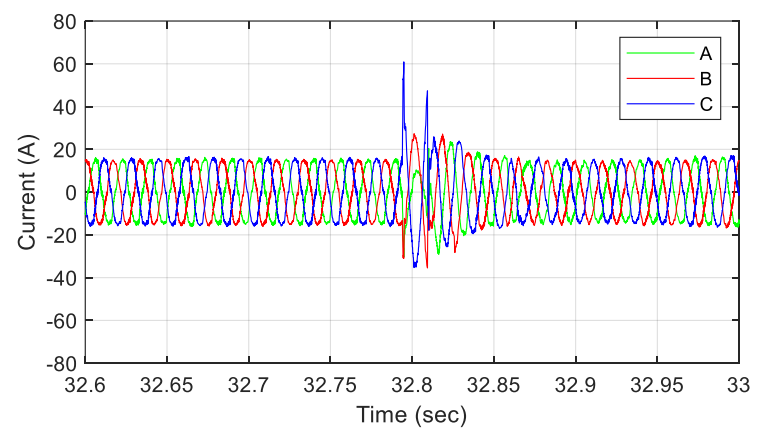
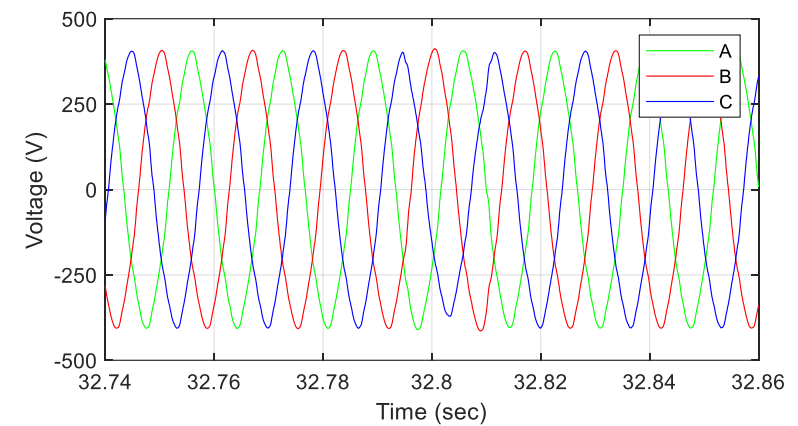
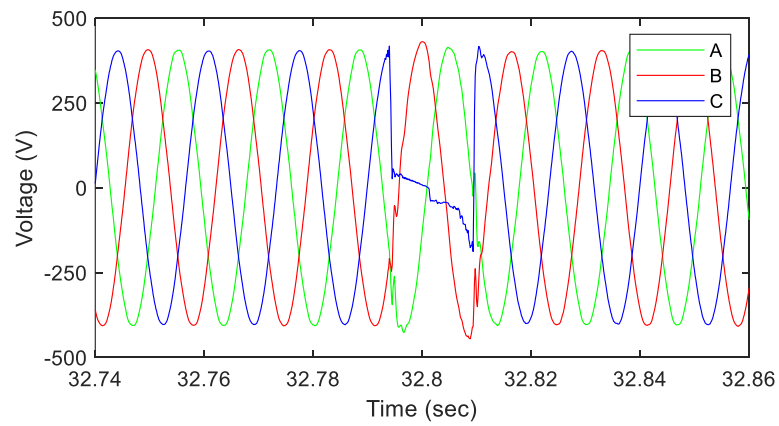
GPS Antennas →



- Analysis of Synchro-Waveforms is the focus in this Tutorial.

Synchro-Waveforms

- Synchro-Waveforms in Example 3:



WMU 1

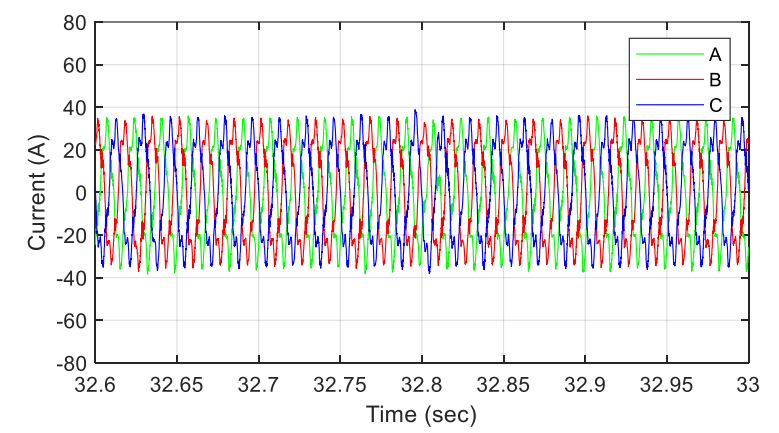
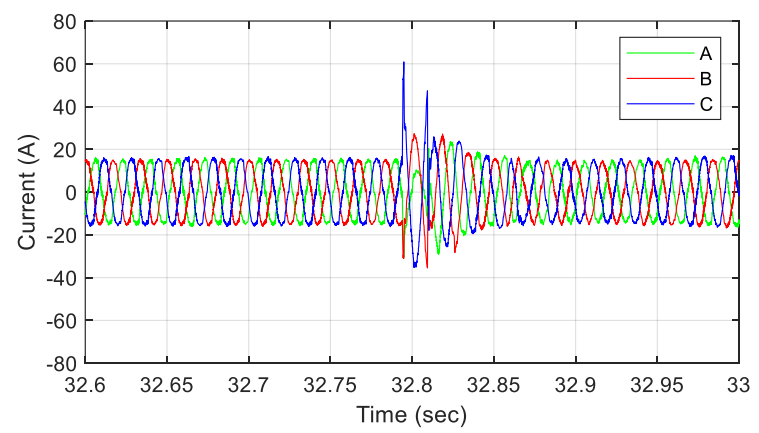
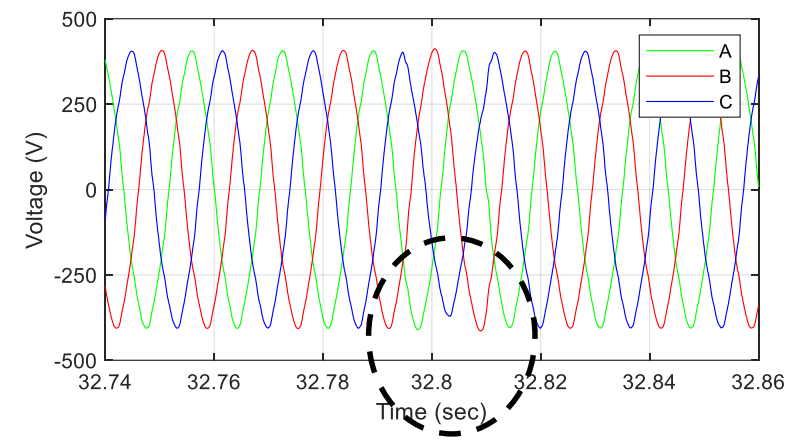
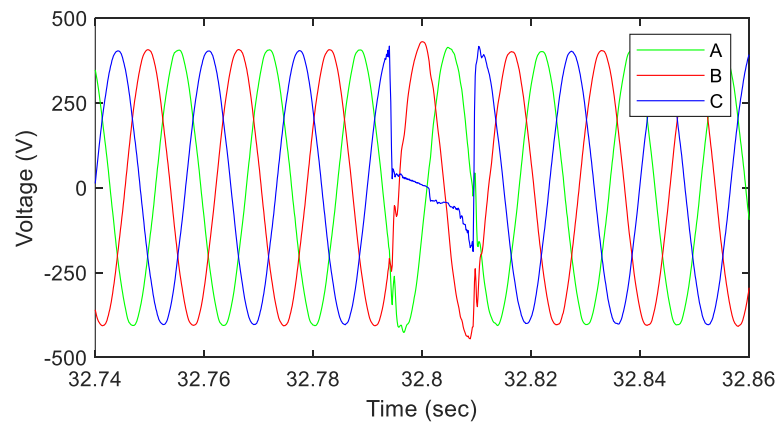


WMU 2

Time Synchronized

Synchro-Waveforms

- Synchro-Waveforms in Example 3:



WMU 1

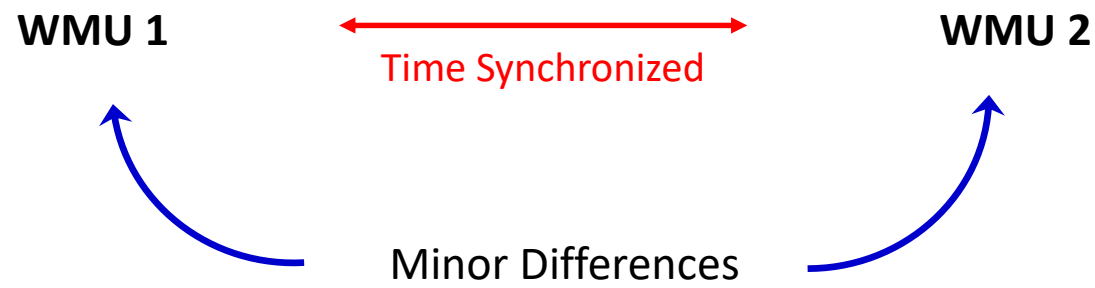
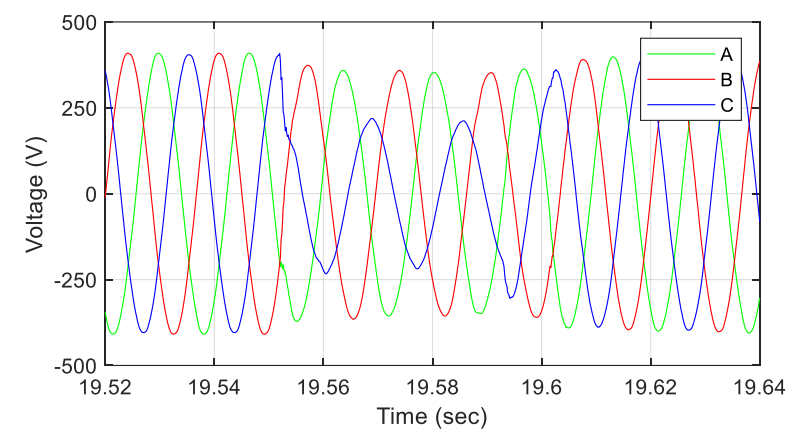
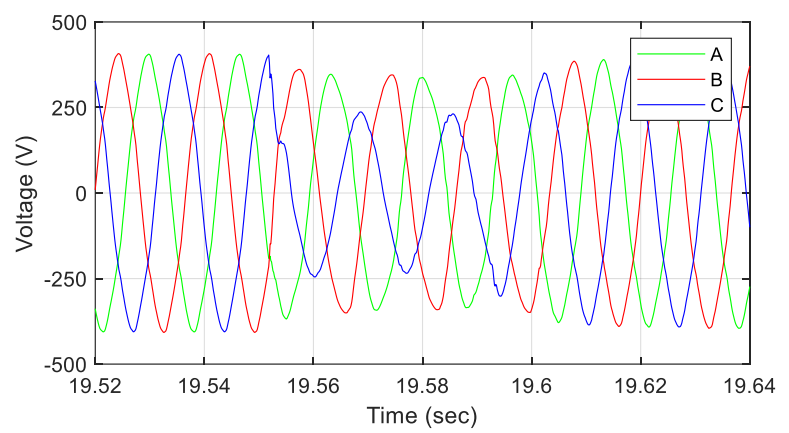


WMU 2

Time Synchronized

Synchro-Waveforms

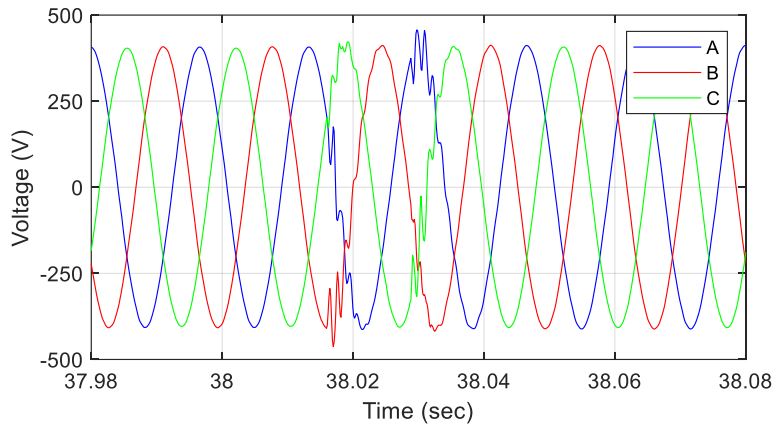
- Another Example - Synchro-Waveforms:



(Event is Likely Far from both WMU 1 and WMU 2)

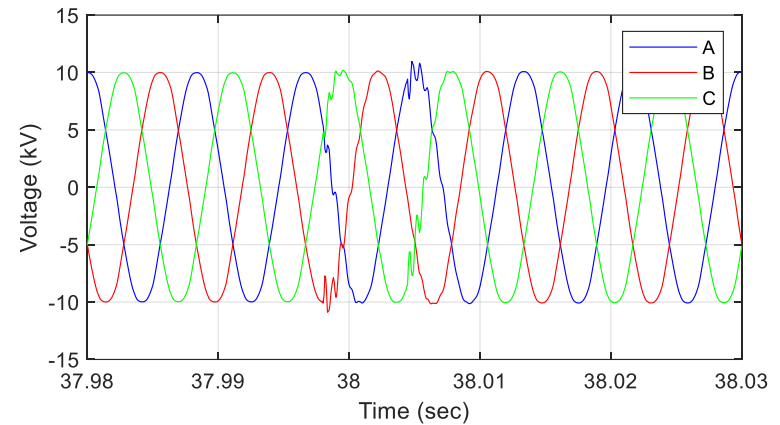
Synchro-Waveforms

- Synchro-Waveforms in the example with **Resonance**:



WMU 1

Time Synchronized



WMU 3

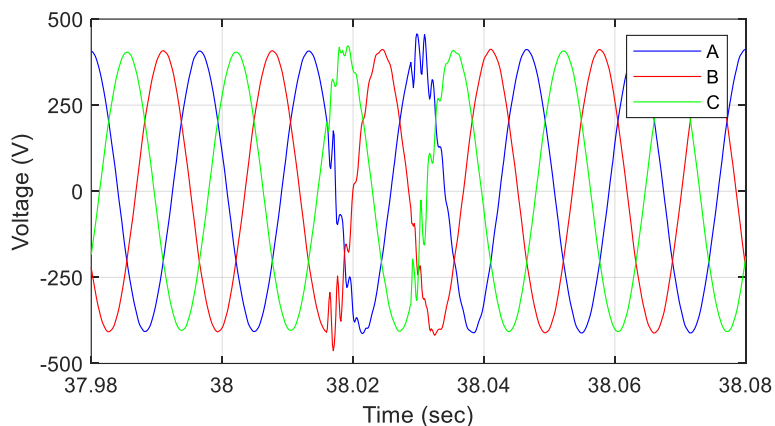
Another Location

(Also, Different Voltage Level)

(System-Wide Sub-Cycle Resonance, Seen at Multiple Substations)

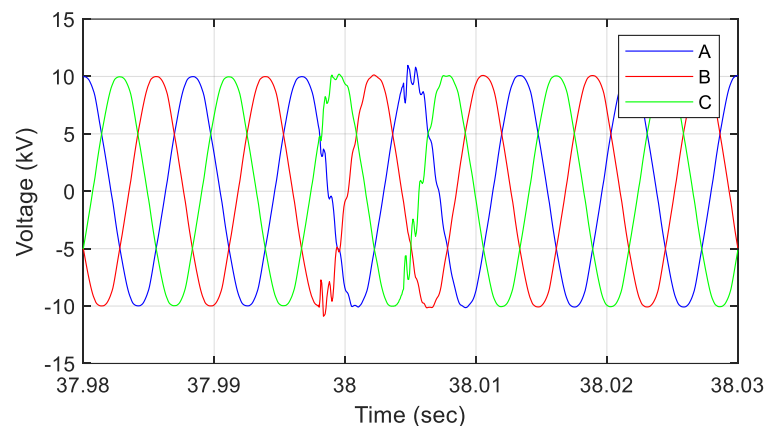
Synchro-Waveforms

- Synchro-Waveforms in the example with **Resonance**:



WMU 1

← Time Synchronized →



WMU 3

- WMUs observe the **same** physical phenomena at **different** locations.

→ **Synchro-Waveform Situational Awareness**

→ **Covering Various Event Signatures (Sub-Cycle, Few-Cycle, etc.)**

Synchro-Waveforms

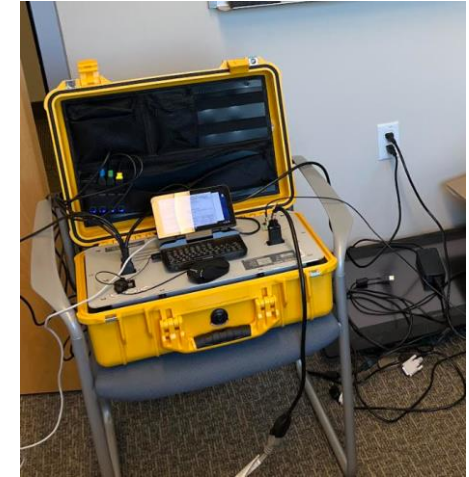
- Field Measurements:



Three-Phase (12.47 kV)



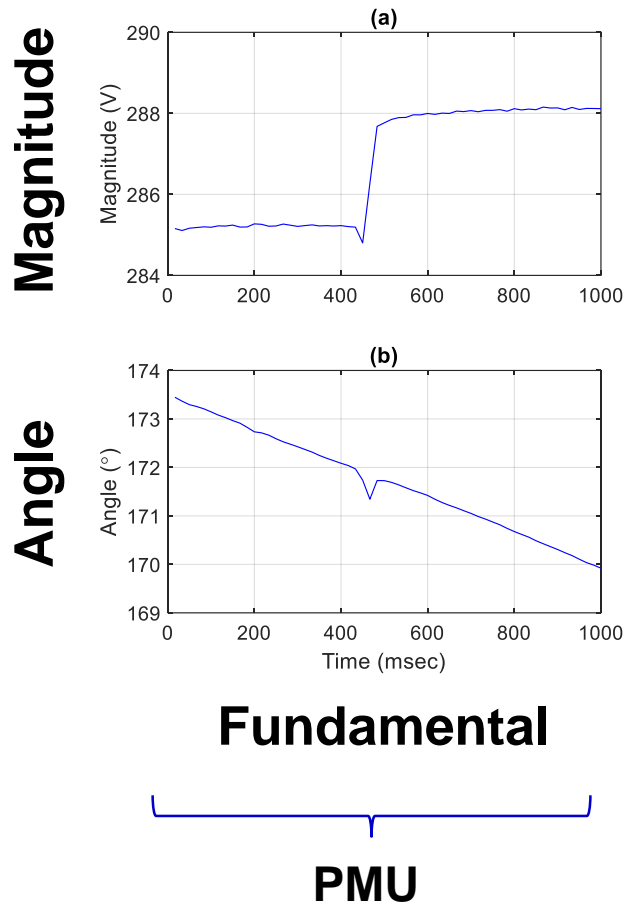
Three-Phase (480 V)



Single-Phase (120 V)

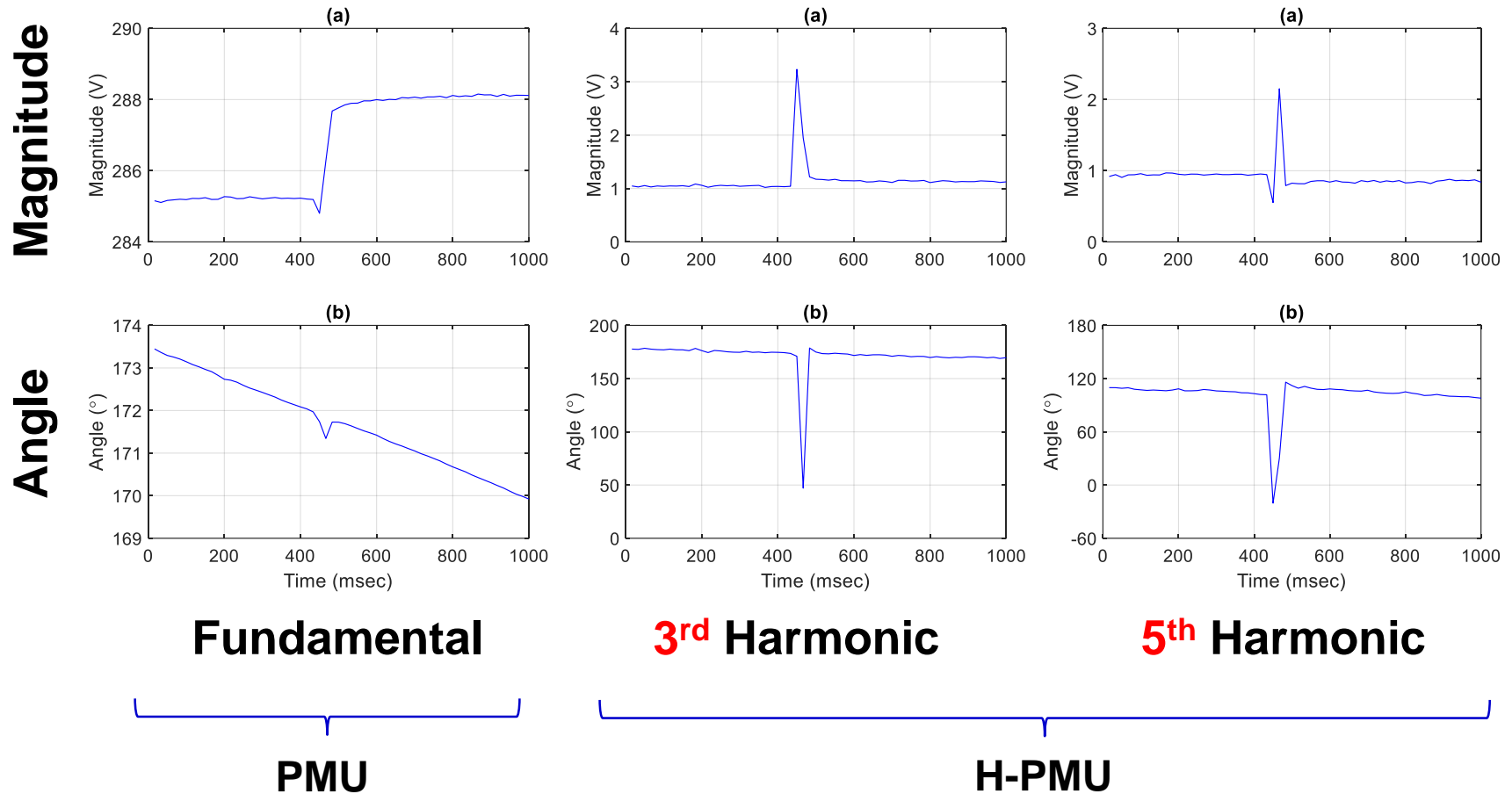
WMUs versus PMUs versus H-PMUs

- **Example:** Phasor Measurements During an **Event:**




WMUs versus PMUs versus H-PMUs

- **Example:** Phasor Measurements During an **Event:**




WMUs versus PMUs versus H-PMUs

- **Note:** Adding *harmonic phasor data* is helpful:
 - **Example:** Improvement in Event Clustering²  13% Improvement

² A. Aligholian and H. Mohsenian-Rad, "GraphPMU: Event Clustering via Graph Representation Learning Using Locationally-Scarce Distribution-Level Fundamental and Harmonic PMU Data," in *IEEE Trans. on Smart Grid*, 2022.

WMUs versus PMUs versus H-PMUs

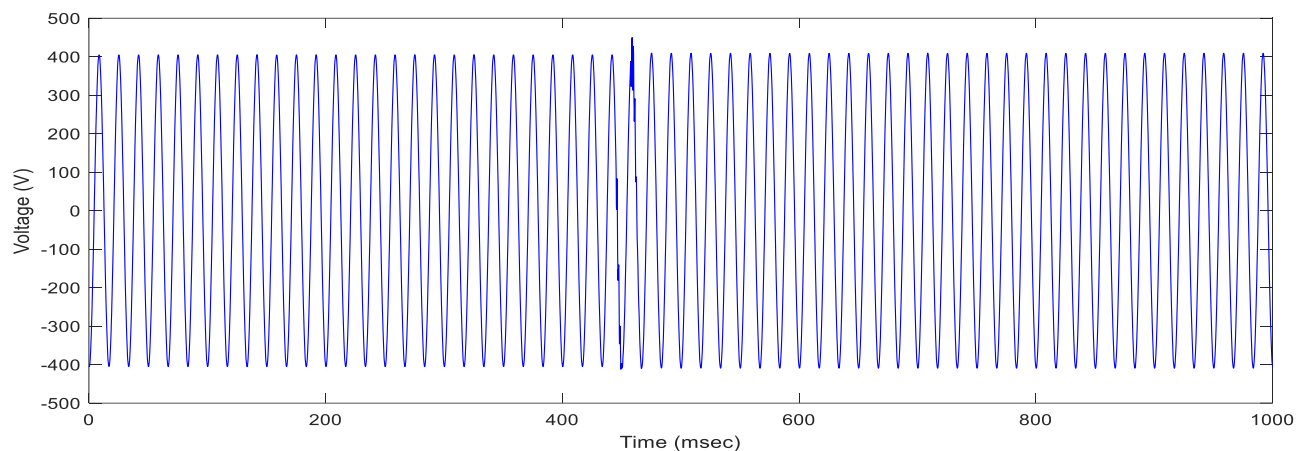
- **Note:** Adding *harmonic phasor data* is helpful:
 - **Example:** Improvement in Event Clustering²  13% Improvement
- **Question:** Why limit ourselves to “phasor” representation?

² A. Aligholian and H. Mohsenian-Rad, "GraphPMU: Event Clustering via Graph Representation Learning Using Locationally-Scarce Distribution-Level Fundamental and Harmonic PMU Data," in *IEEE Trans. on Smart Grid*, 2022.

WMUs versus PMUs versus H-PMUs

- **Note:** Adding *harmonic phasor data* is helpful:
 - **Example:** Improvement in Event Clustering² → 13% Improvement
- **Question:** Why limit ourselves to “phasor” representation?

**Waveform Data
(Same Period)**

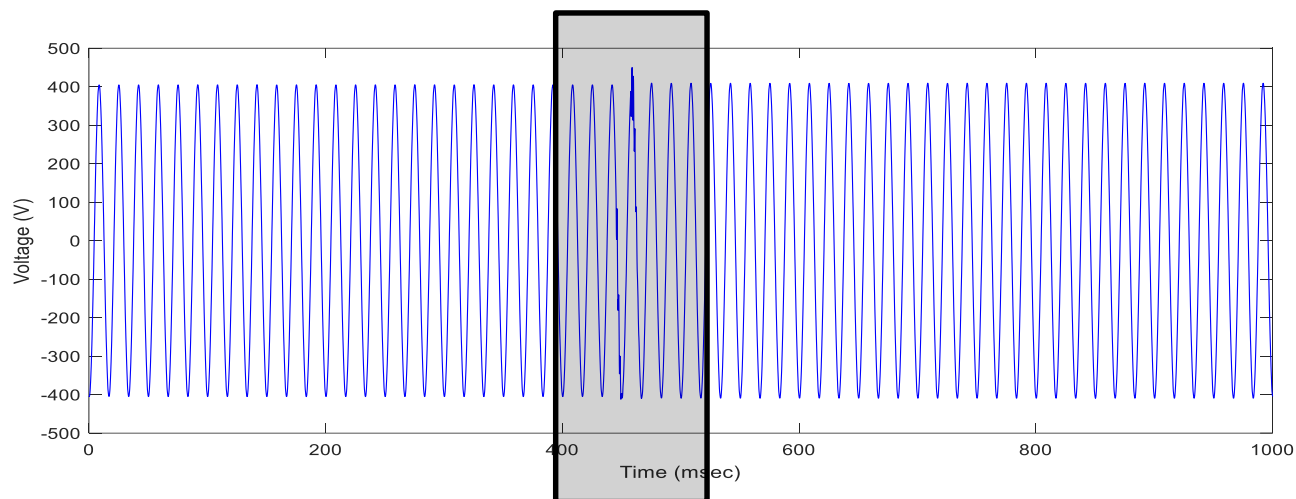


² A. Aligholian and H. Mohsenian-Rad, "GraphPMU: Event Clustering via Graph Representation Learning Using Locationally-Scarce Distribution-Level Fundamental and Harmonic PMU Data," in *IEEE Trans. on Smart Grid*, 2022.

WMUs versus PMUs versus H-PMUs

- **Note:** Adding *harmonic phasor data* is helpful:
 - **Example:** Improvement in Event Clustering² → 13% Improvement
- **Question:** Why limit ourselves to “phasor” representation?

**Waveform Data
(Same Period)**



² A. Aligholian and H. Mohsenian-Rad, "GraphPMU: Event Clustering via Graph Representation Learning Using Locationally-Scarce Distribution-Level Fundamental and Harmonic PMU Data," in *IEEE Trans. on Smart Grid*, 2022.

Synchro-Waveform Data Analysis

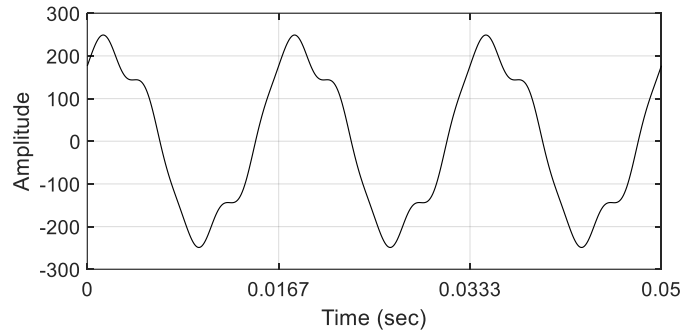
- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Synchro-Waveform Data Analysis

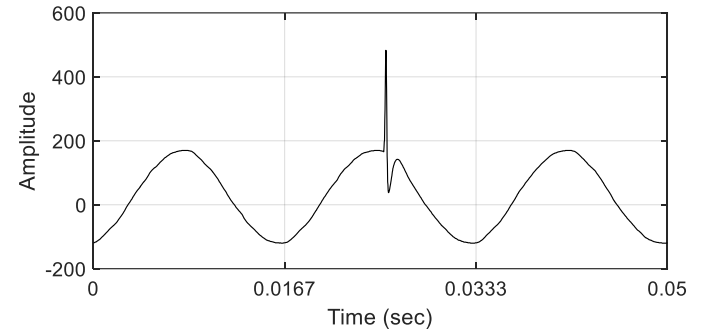
- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Event Detection

- Let's distinguish two cases:



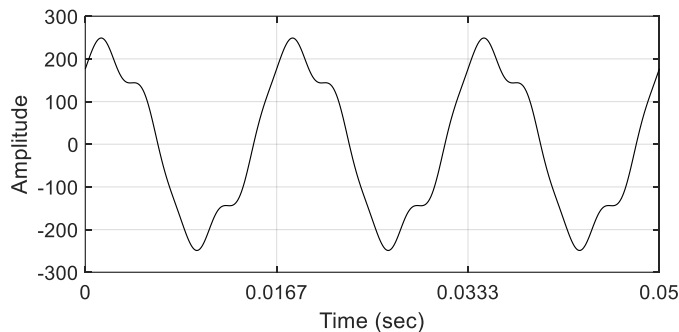
Harmonic Distortions



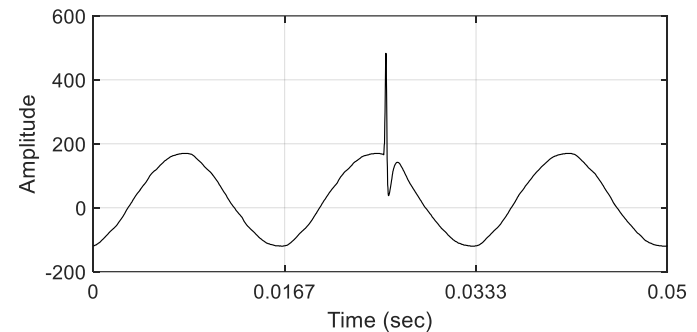
Event

Event Detection

- Let's distinguish two cases:



Harmonic Distortions



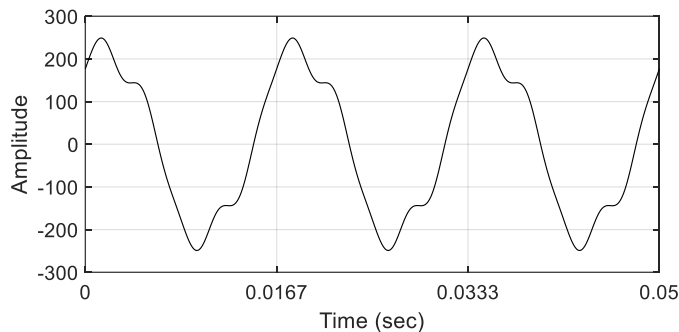
Event

(Steady-State Analysis³ \rightarrow H-PMUs)

³ F. Ahmadi and H. Mohsenian-Rad, "A Physics-Aware MIQP Approach to Harmonic State Estimation in Low-Observable Power Distribution Systems Using Harmonic Phasor Measurement Units," in *IEEE Trans. on Smart Grid*, Sept, 2022.

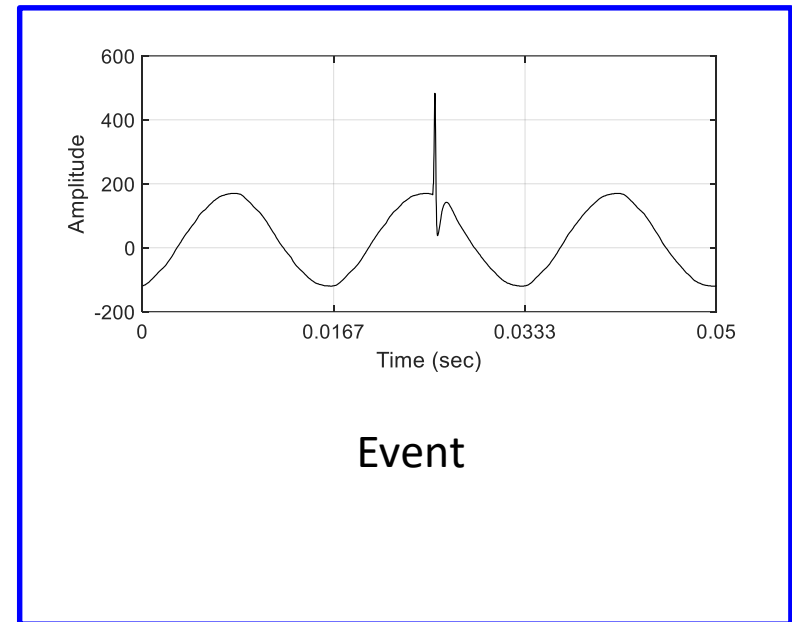
Event Detection

- Let's distinguish two cases:



Harmonic Distortions

(Steady-State Analysis³ \rightarrow H-PMUs)



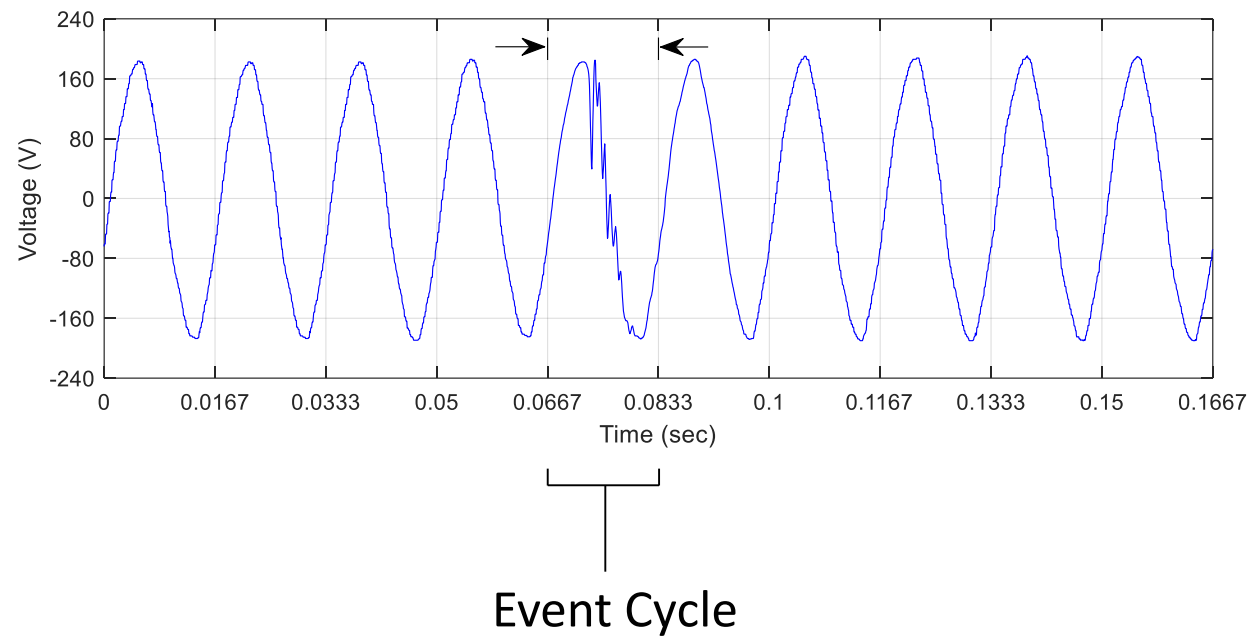
Event

Our Focus in This Section

³ F. Ahmadi and H. Mohsenian-Rad, "A Physics-Aware MIQP Approach to Harmonic State Estimation in Low-Observable Power Distribution Systems Using Harmonic Phasor Measurement Units," in *IEEE Trans. on Smart Grid*, Sept, 2022.

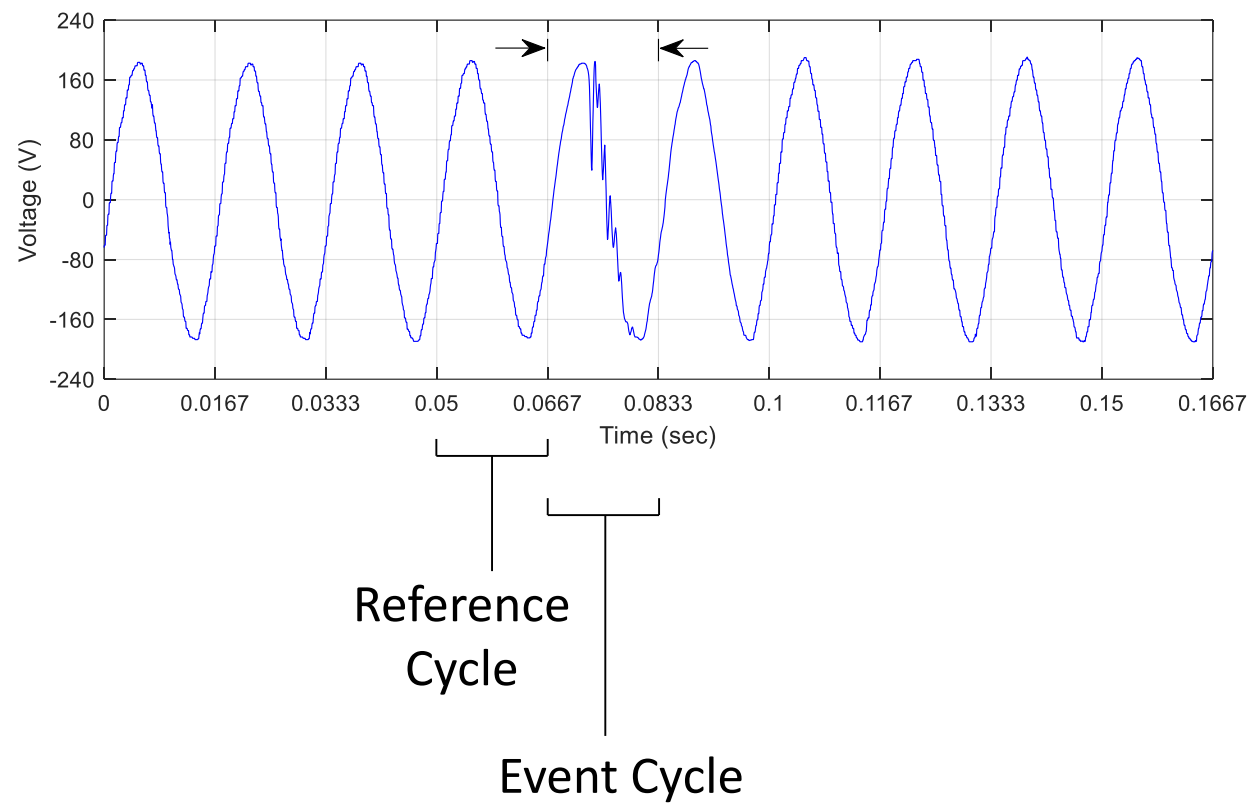
Event Detection

- Event-triggered waveform capture:



Event Detection

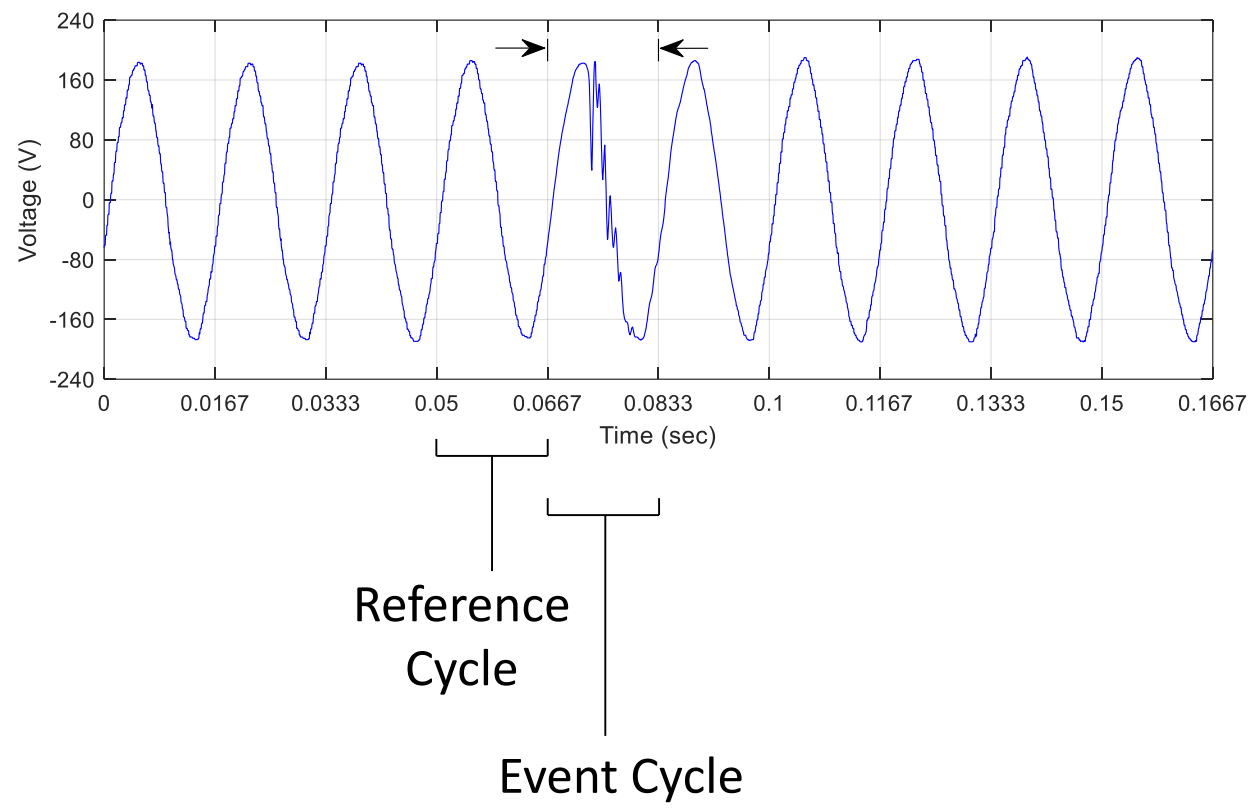
- In practice, it is common to simply *compare two consecutive cycles*:



Event Detection

- In practice, it is common to simply *compare two consecutive cycles*:

What Metric?

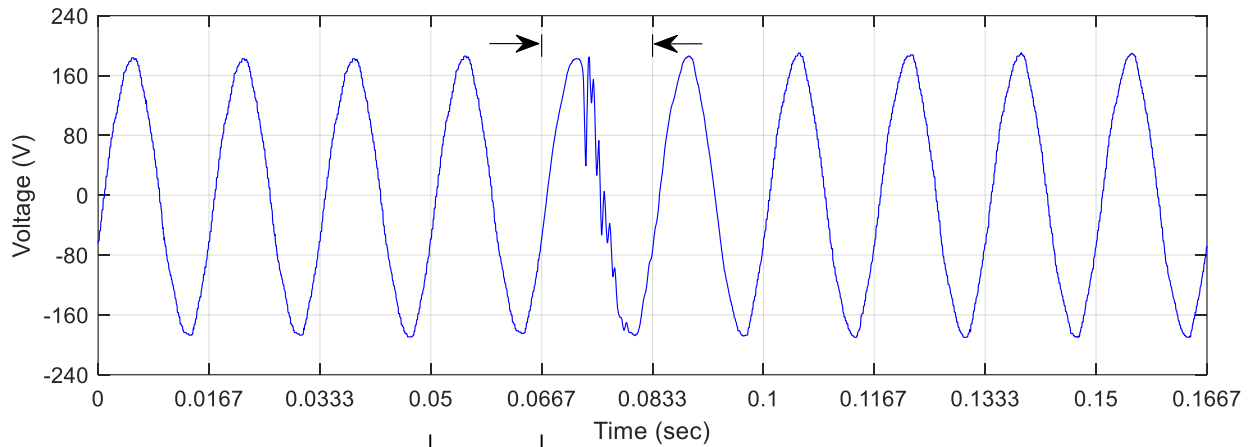


- Different ways to compare two cycles of waveforms¹:
 - Comparing THD
 - Comparing RMS
 - Point-to-Point Comparison
 - Comparing Sub-Cycle RMS
 - Differential Waveform
 - Neutral Current Waveform
 - Other Factors and Methods

- Different ways to compare two cycles of waveforms¹:
 - Comparing THD ←
 - Comparing RMS
 - Point-to-Point Comparison
 - Comparing Sub-Cycle RMS
 - Differential Waveform ←
 - Neutral Current Waveform ←
 - Other Factors and Methods

Comparing THD

- Compare two consecutive waveform cycles based on their THD values.



THD < 1%

THD = Total Harmonic Distortion

$$|\Delta\text{THD}| \geq \alpha_{\text{THD}}$$

THD = 16%

Differential Waveform

- It works based on obtaining the following *differential waveform*:

$$\Delta x(t) = x(t) - x(t - NT).$$

where

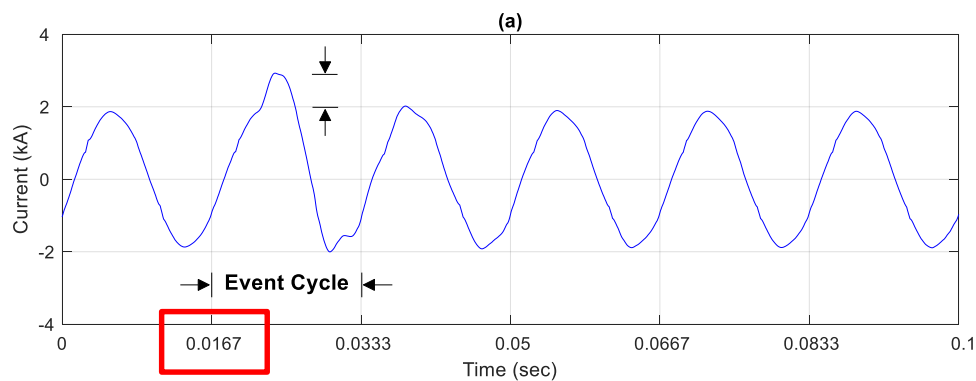
$x(t)$ is the measured current waveform or voltage waveform;
 T is the waveform interval; and
 N is a small integer number, e.g., 1, 2, 3, 4, or 5.

- We can detect an event based on the characteristics of $\Delta x(t)$.

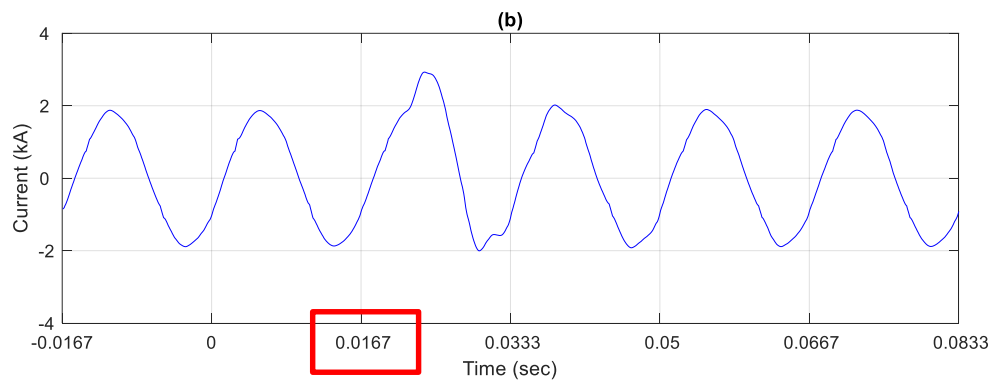
Differential Waveform

- Consider the current waveform measurements below:

$x(t)$

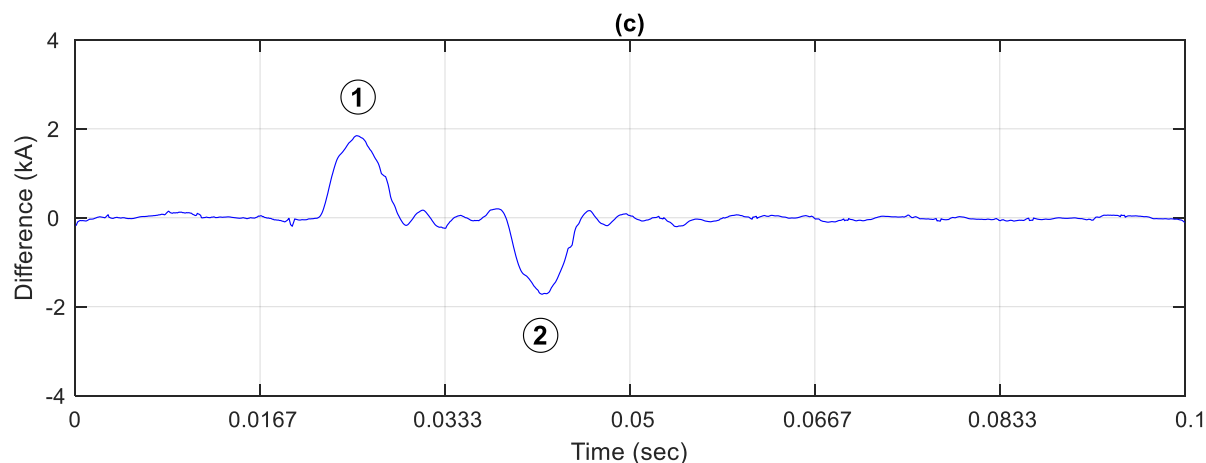


$x(t - T)$



Differential Waveform

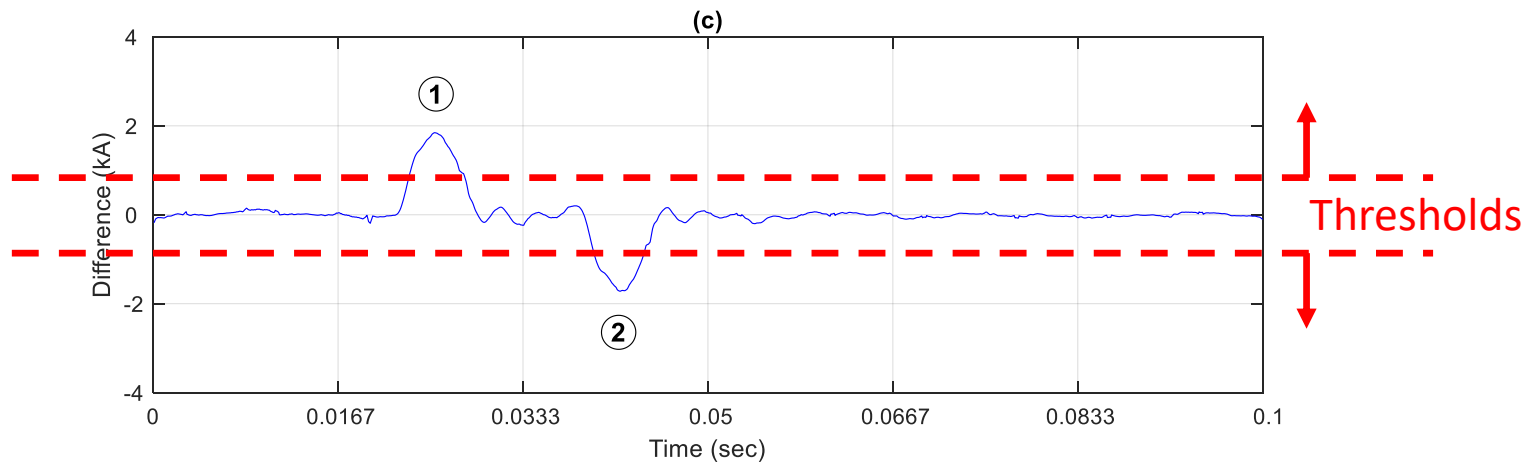
- The differential waveform is obtained as:



- We can see that the event has created two distinct blips in the differential waveform, which are denoted by ① and ② .
- Note that *both* of them are associated with the *same* event.

Differential Waveform

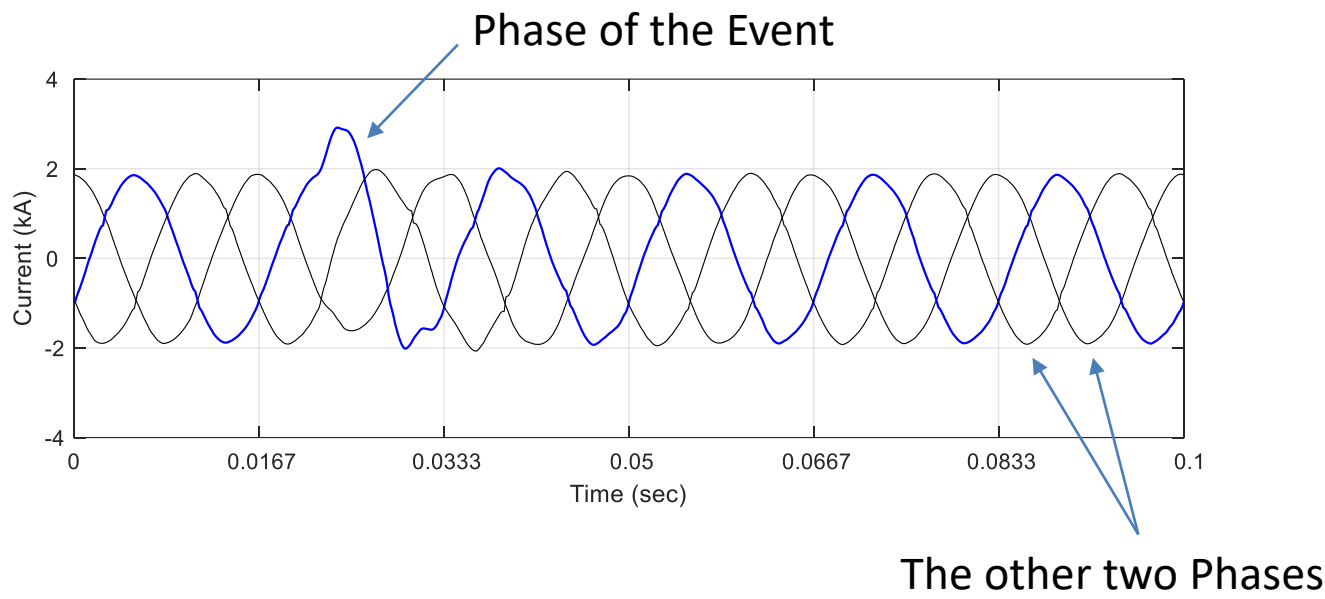
- The differential waveform is obtained as:



- We can see that the event has created two distinct blips in the differential waveform, which are denoted by ① and ②.
- Note that *both* of them are associated with the *same* event.

Neutral Current Waveform

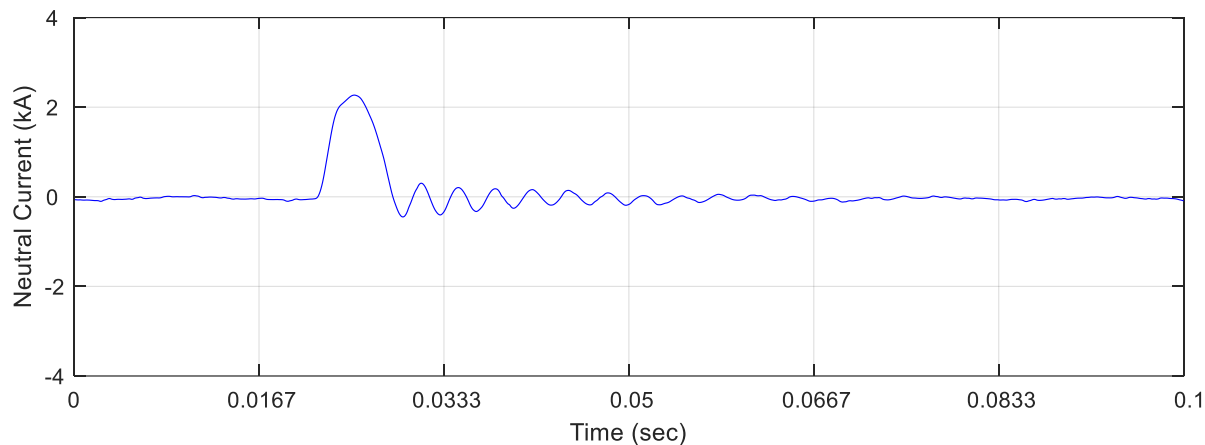
- Consider the following three-phase current waveform measurements:



Neutral Current Waveform

- The neutral current is obtained as:

$$i_N(t) = i_A(t) + i_B(t) + i_C(t).$$

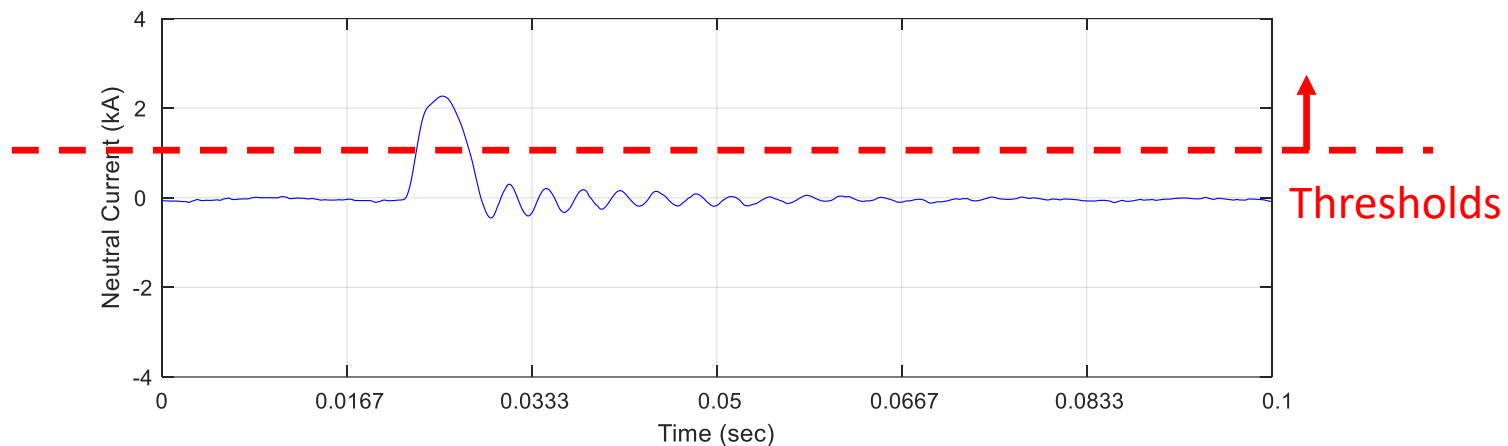


- **Note:** No second blip, unlike in the differential waveforms.

Neutral Current Waveform

- The neutral current is obtained as:

$$i_N(t) = i_A(t) + i_B(t) + i_C(t).$$







- **Note:** No second blip, unlike in the differential waveforms.

Event Detection – Multiple Waveforms

- We may also try to *simultaneously* check multiple waveforms.
- For example, suppose two WMUs collect the following waveforms:
 - Voltage at WMU 1: $v_1(t)$
 - Current at WMU 1: $i_1(t)$
 - Voltage at WMU 2: $v_2(t)$
 - Current at WMU 2: $i_2(t)$

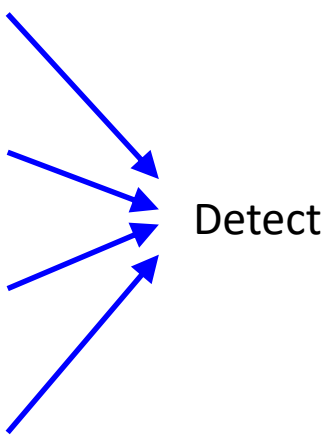
Event Detection – Multiple Waveforms

- We may also try to *simultaneously* check multiple waveforms.
- For example, suppose two WMUs collect the following waveforms:
 - Voltage at WMU 1: $v_1(t)$  Detect
 - Current at WMU 1: $i_1(t)$  Detect
 - Voltage at WMU 2: $v_2(t)$  Detect
 - Current at WMU 2: $i_2(t)$  Detect

We can look for event in *each* waveform.

Event Detection – Multiple Waveforms

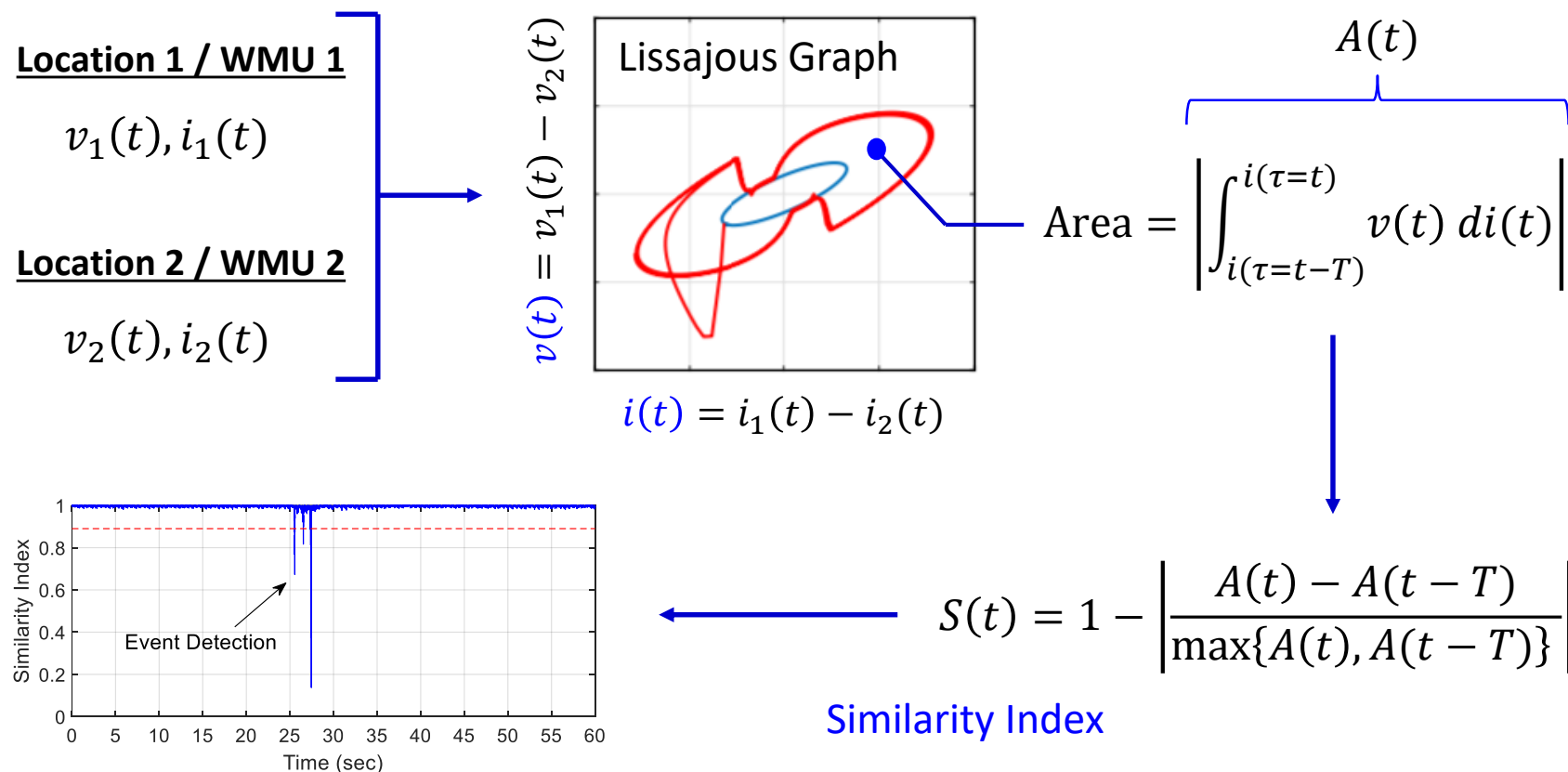
- We may also try to *simultaneously* check multiple waveforms.
- For example, suppose two WMUs collect the following waveforms:

- Voltage at WMU 1: $v_1(t)$
 - Current at WMU 1: $i_1(t)$
 - Voltage at WMU 2: $v_2(t)$
 - Current at WMU 2: $i_2(t)$
- 
- Detect

We can look for event in *all* waveforms.

Event Detection – Multiple Waveforms

- Graphical Metrics⁴:



⁴ M. Izadi and H. Mohsenian-Rad, "Characterizing synchronized Lissajous curves to scrutinize power distribution synchro-waveform measurements," in *IEEE Trans. on Power Systems*, vol. 36, no. 5, pp. 4880-4884, Sept 2021.

Synchro-Waveform Data Analysis

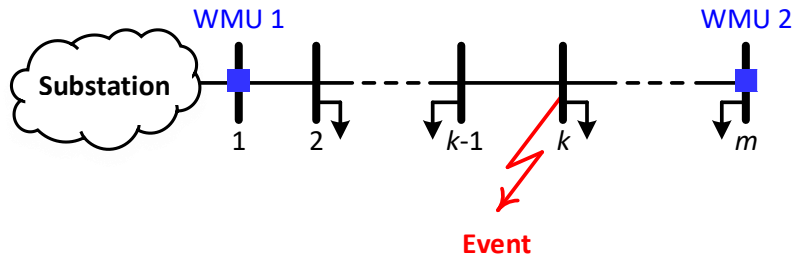
- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Synchro-Waveform Data Analysis

- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

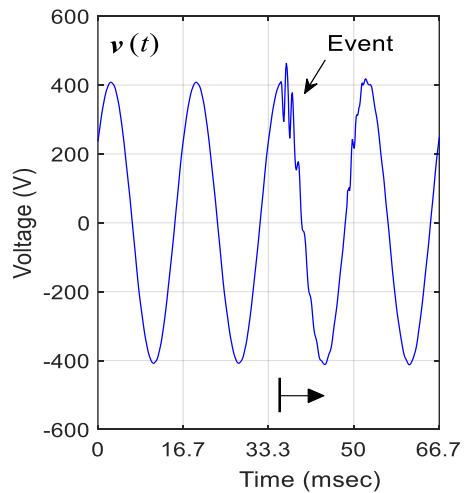
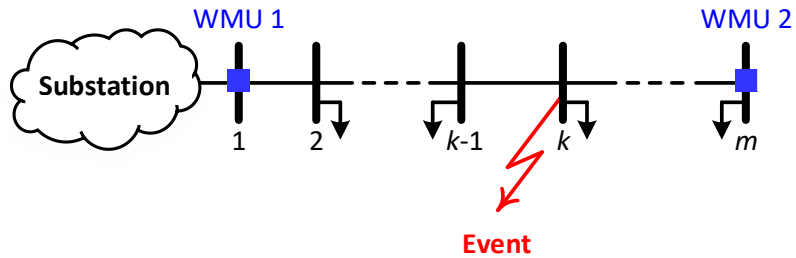
Event Location Identification

- Finding the Cause of Transient Events:



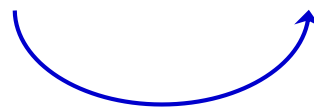
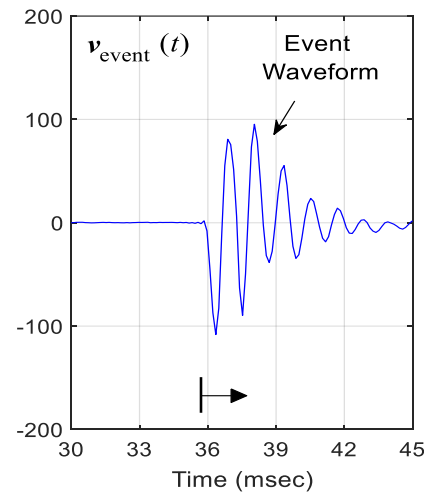
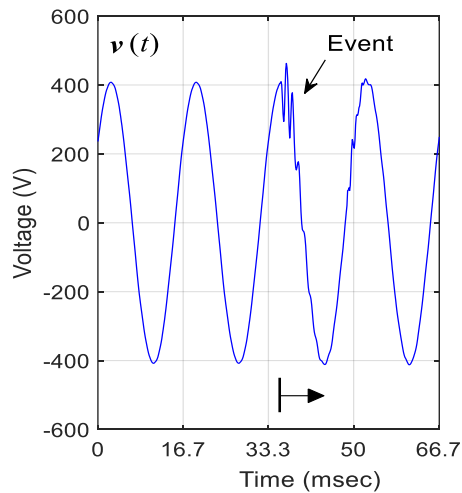
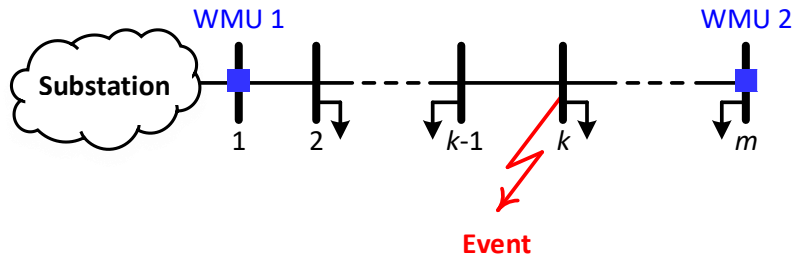
Event Location Identification

- Finding the Cause of Transient Events:



Event Location Identification

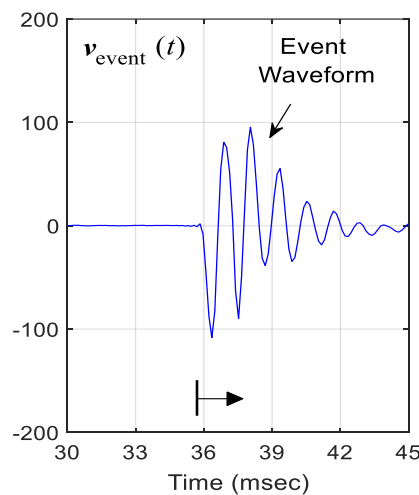
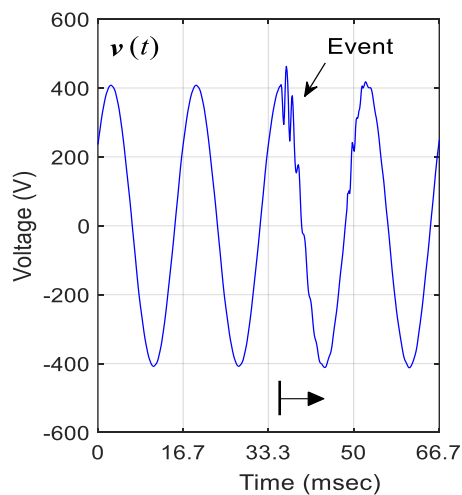
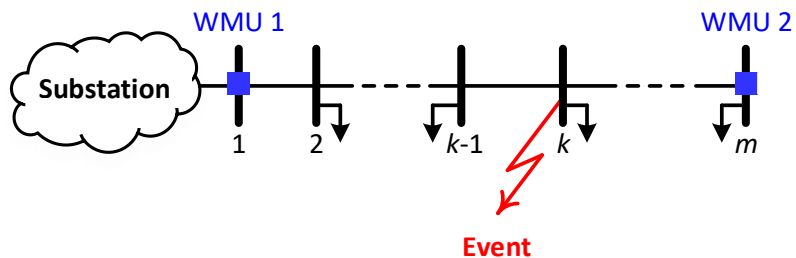
- Finding the Cause of Transient Events:



Extract $v_{\text{event}}(t)$

Event Location Identification

- Finding the Cause of Transient Events:



Extract $v_{event}(t)$

Modal Analysis

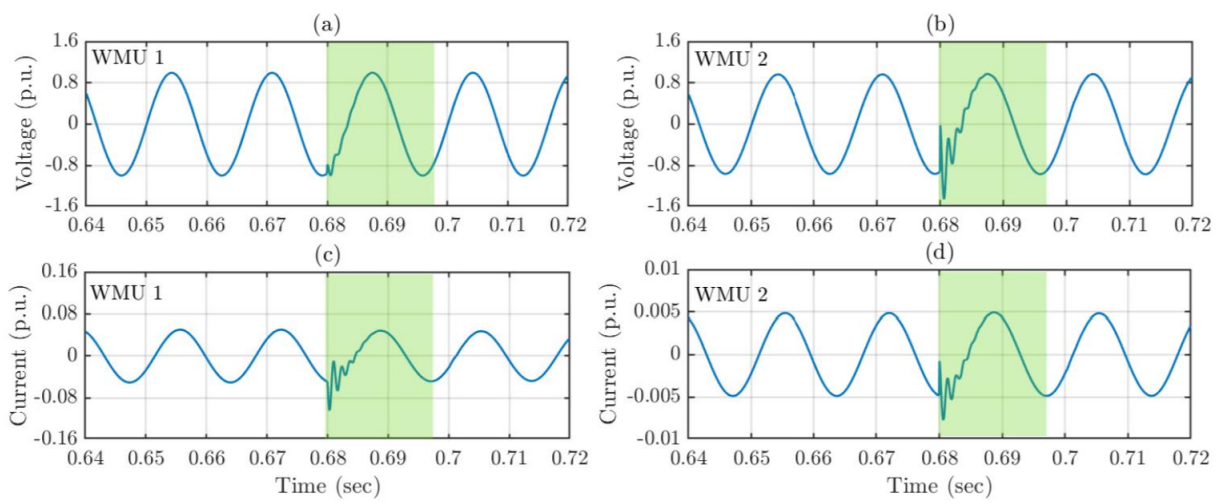
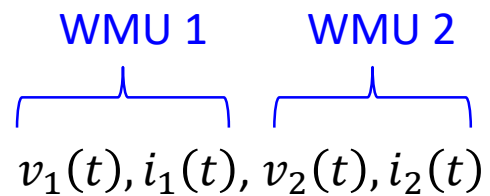
Number of Modes

$$\sum_{p=1}^P A_{p,m} e^{\sigma_p t} \cos(2\pi f_p t + \theta_p)$$

Damping Sinusoidal Modes
($f_p, \sigma_p, A_{p,m}, \theta_p$)

Event Location Identification

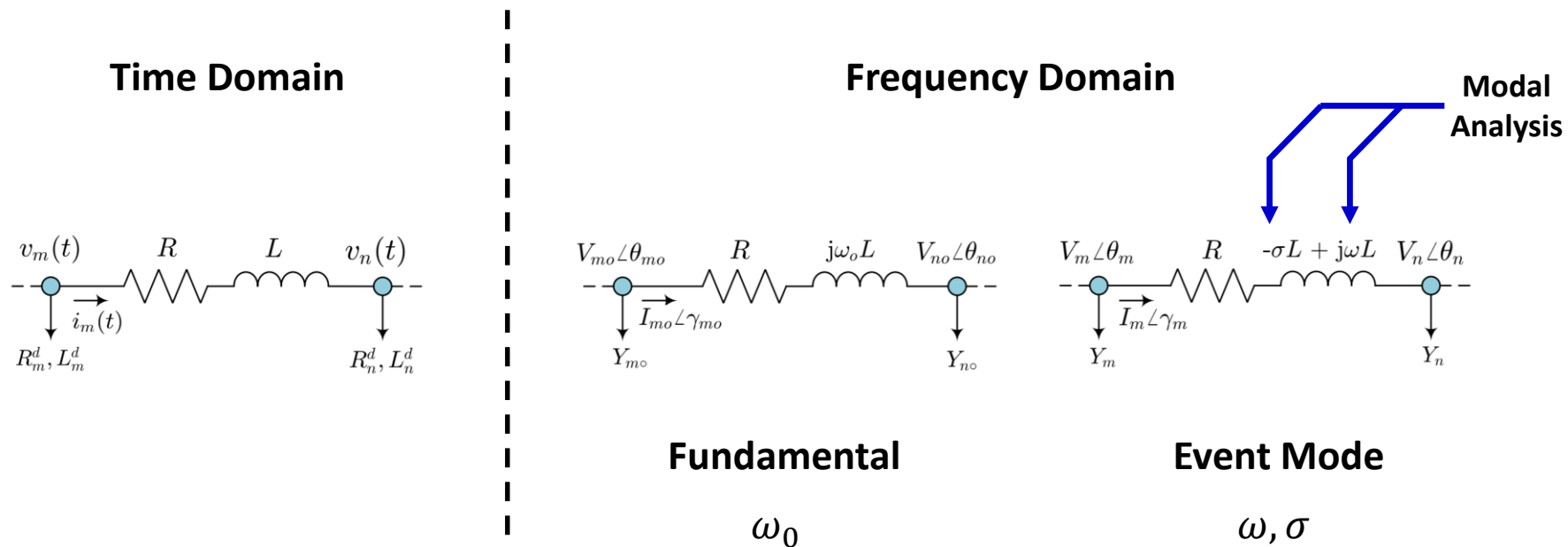
- Multi-Signal Modal Analysis:



WMU	Signal	Frequency (Hz)	Damping Rate (Hz)	Magnitude (p.u.)	Phase Angle (deg.)
1	Voltage	60.00 / 747.72	0.00 / -624.30	0.98 / 0.20	0.00 / 0.00
	Current			0.04 / 0.06	-25.19 / 82.43
2	Voltage	60.00 / 747.72	0.00 / -624.30	0.96 / 0.92	-0.49 / -1.07
	Current			0.004 / 0.004	-25.96 / -3.23

* The two most dominant modes are separated with a slash.

Event Location Identification



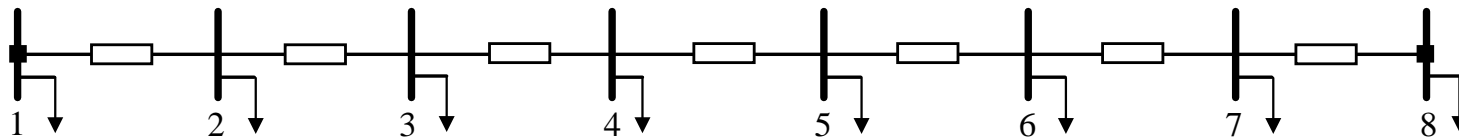
- Accordingly, we can solve the circuit in “event mode”.
 - This means solving the circuit based on ω, σ (instead of over ω_0)

Event Location Identification

- Circuit model under the event mode⁵:

WMU 1
 $\omega, \sigma, A_1, \theta_1$

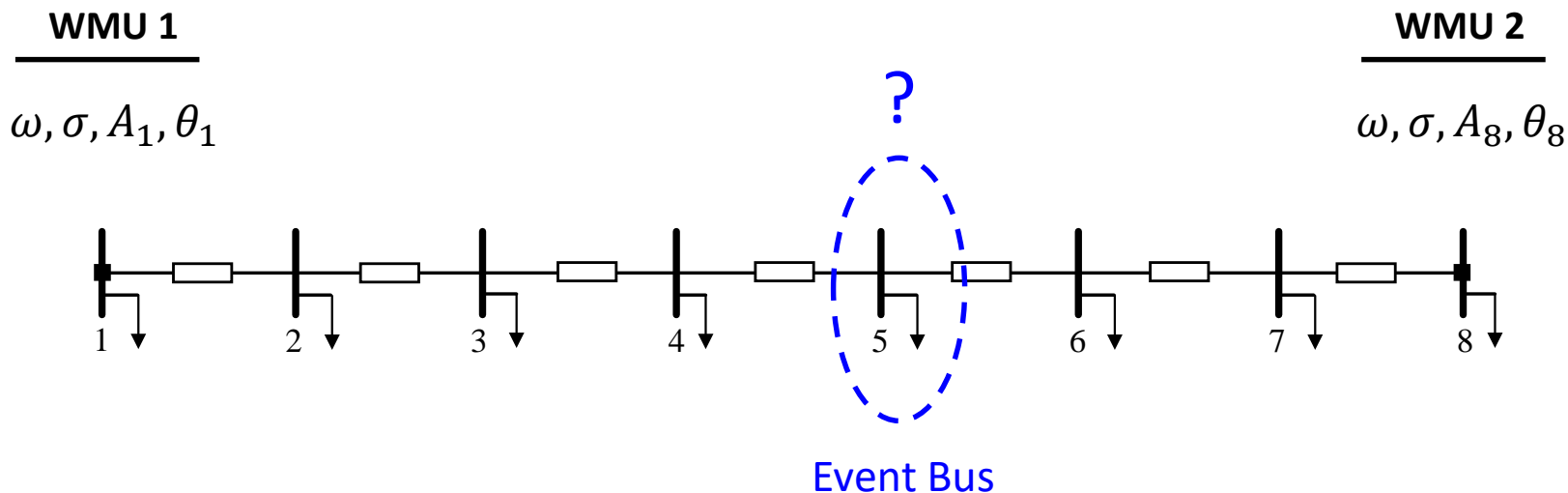
WMU 2
 $\omega, \sigma, A_8, \theta_8$



⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- Circuit model under the event mode⁵:



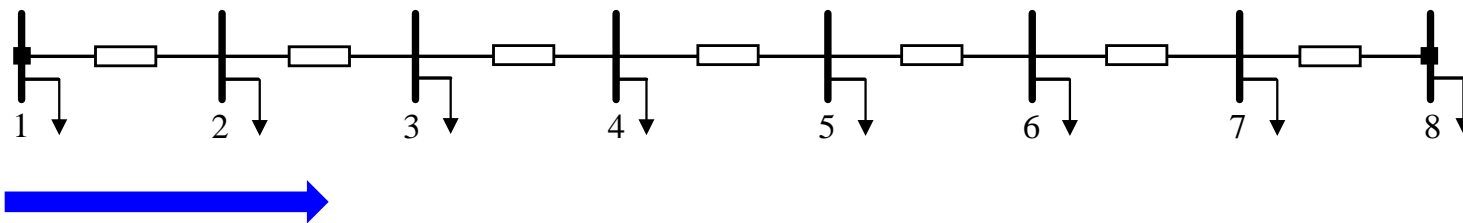
⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- Circuit model under the event mode⁵:

WMU 1
 $\omega, \sigma, A_1, \theta_1$

WMU 2
 $\omega, \sigma, A_8, \theta_8$



Step 1: Forward Sweep (Event Mode)

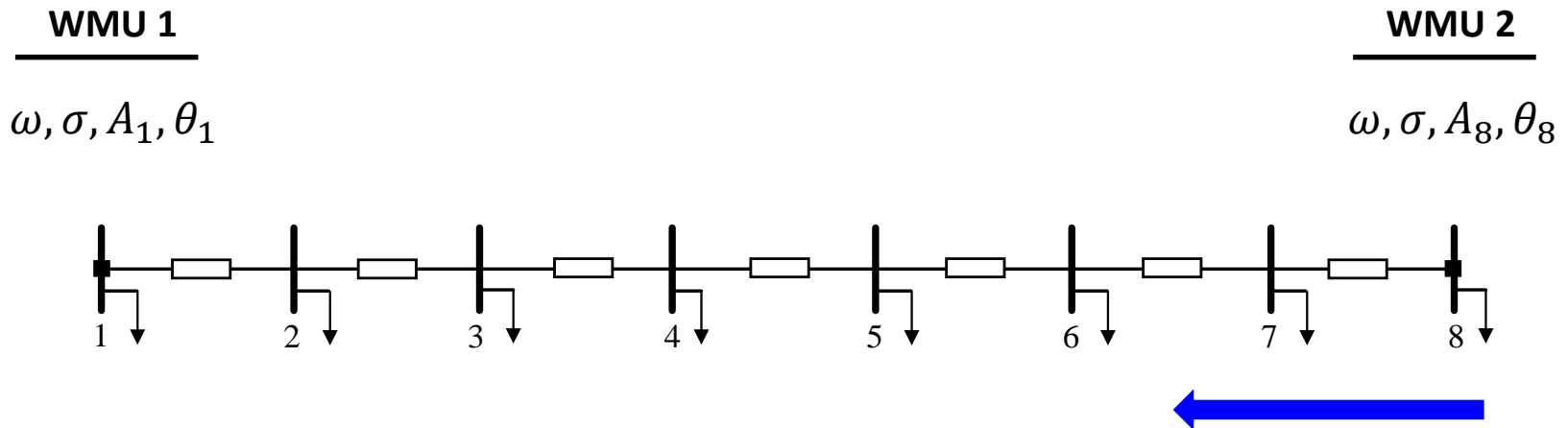
$$V_1^f, V_2^f, \dots, V_7^f, V_8^f$$

(Phasors in Event Mode; *not* in Fundamental Mode)

⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- Circuit model under the event mode⁵:



Step 2: Backward Sweep (Event Mode)

$$V_1^b, V_2^b, \dots, V_7^b, V_8^b$$

(Phasors in Event Mode; *not* in Fundamental Mode)

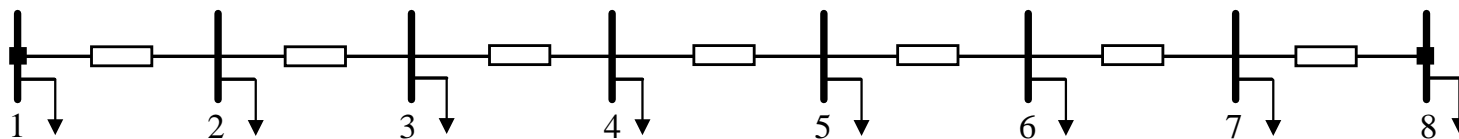
⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- Circuit model under the event mode⁵:

WMU 1
 $\omega, \sigma, A_1, \theta_1$

WMU 2
 $\omega, \sigma, A_8, \theta_8$



Step 3:

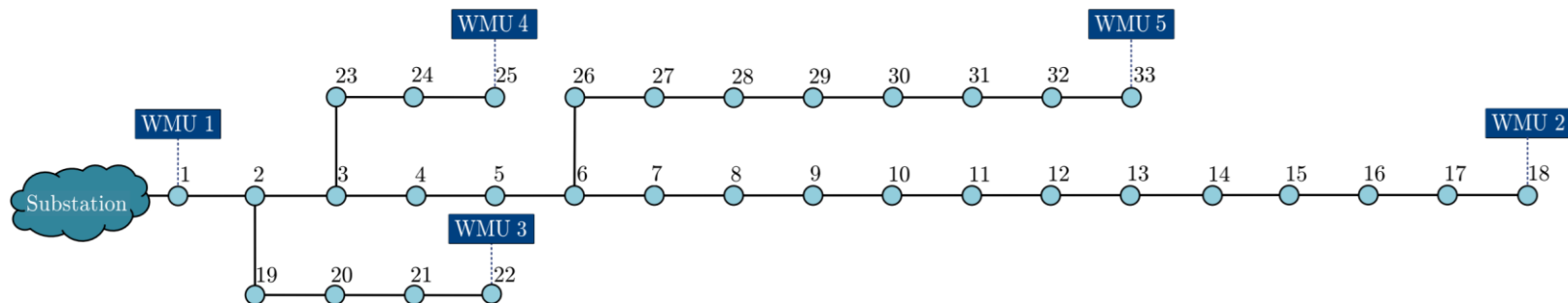
$$k^* = \underset{i}{\operatorname{argmin}} \Psi_i \quad \text{where} \quad \Psi_i = |V_i^f - V_i^b|, \quad i = 1, \dots, 8.$$

↑
Discrepancy

⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- We may have more than two WMUs available:

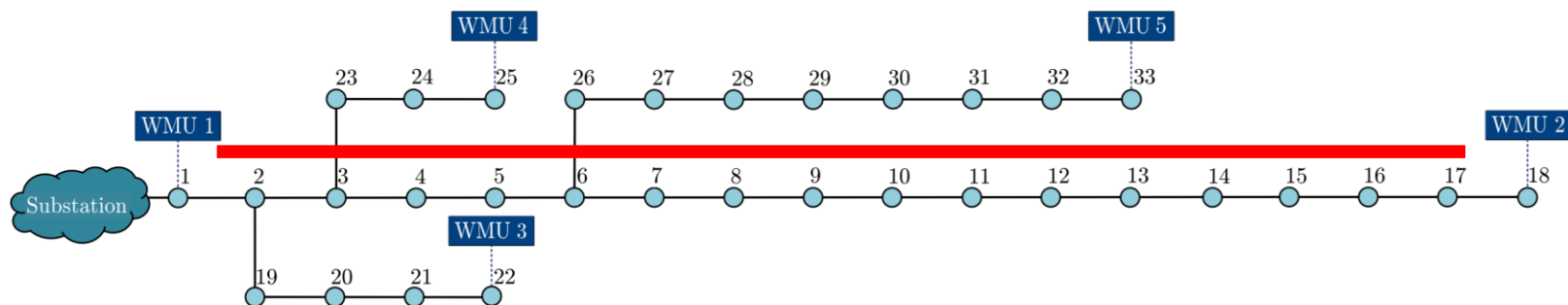


- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

Event Location Identification

- We may have more than two WMUs available:

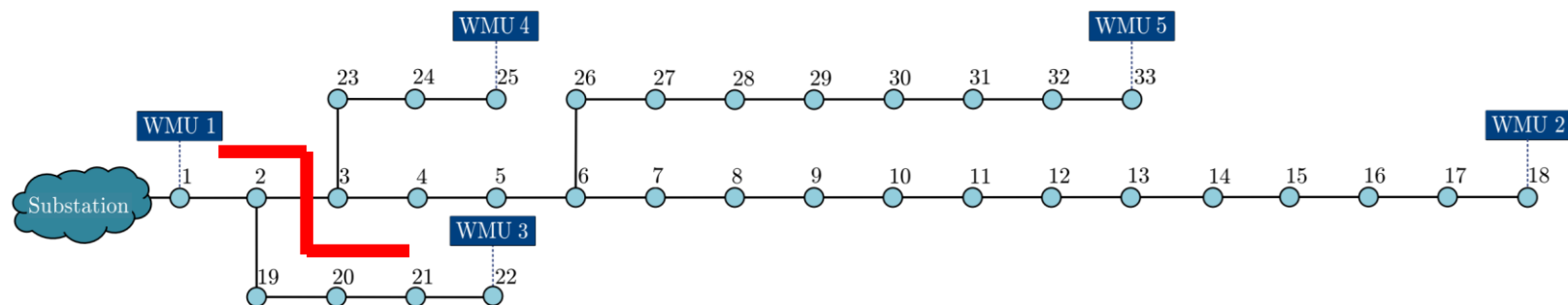


- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

Event Location Identification

- We may have more than two WMUs available:

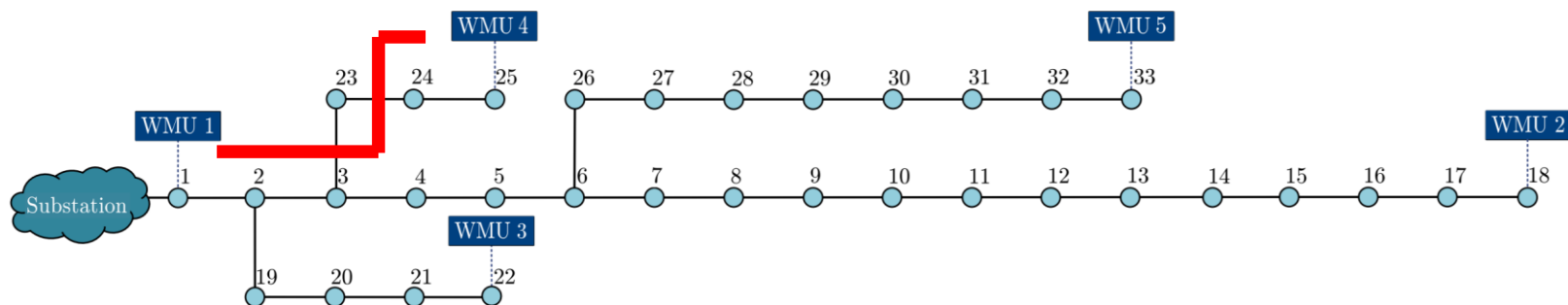


- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

Event Location Identification

- We may have more than two WMUs available:

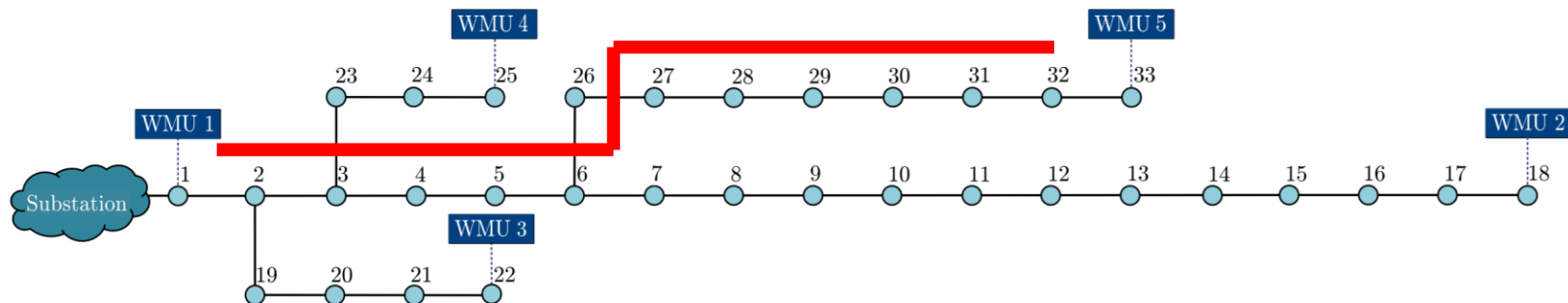


- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

Event Location Identification

- We may have more than two WMUs available:

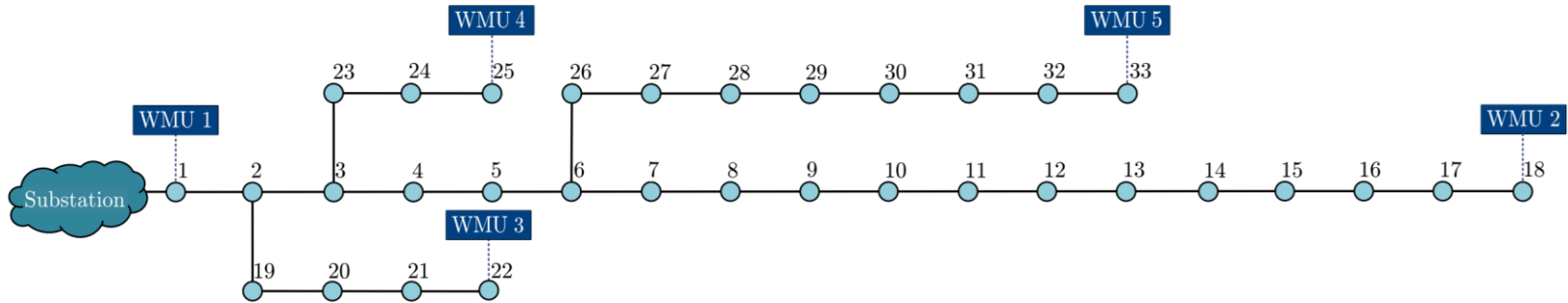


- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

Event Location Identification

- We may have more than two WMUs available:



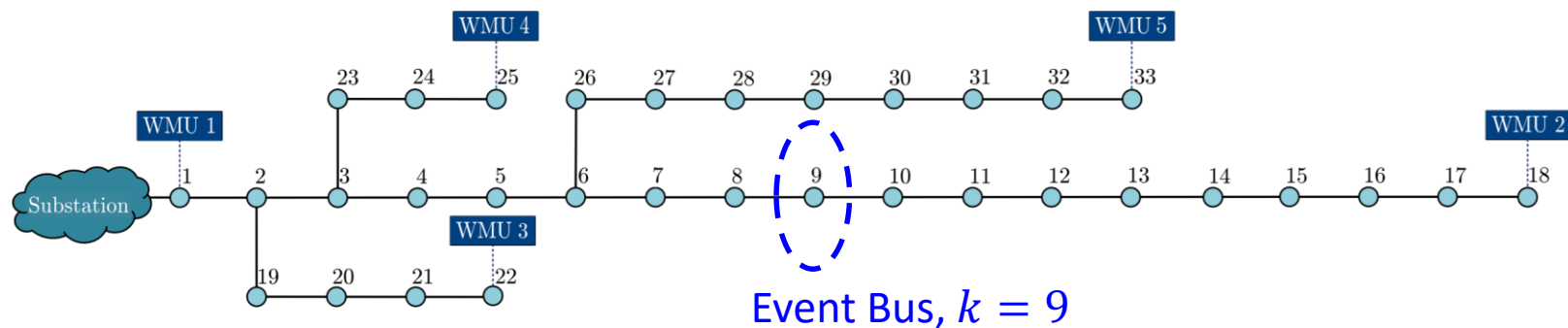
- Several Options:

- 1) WMU 1 and WMU 2
- 2) WMU 1 and WMU 3
- 3) WMU 1 and WMU 4
- 4) WMU 1 and WMU 5

$$k^* = \underset{i}{\operatorname{argmin}} \sum_{s=2}^5 \Psi_i^{1,s}$$

Event Location Identification

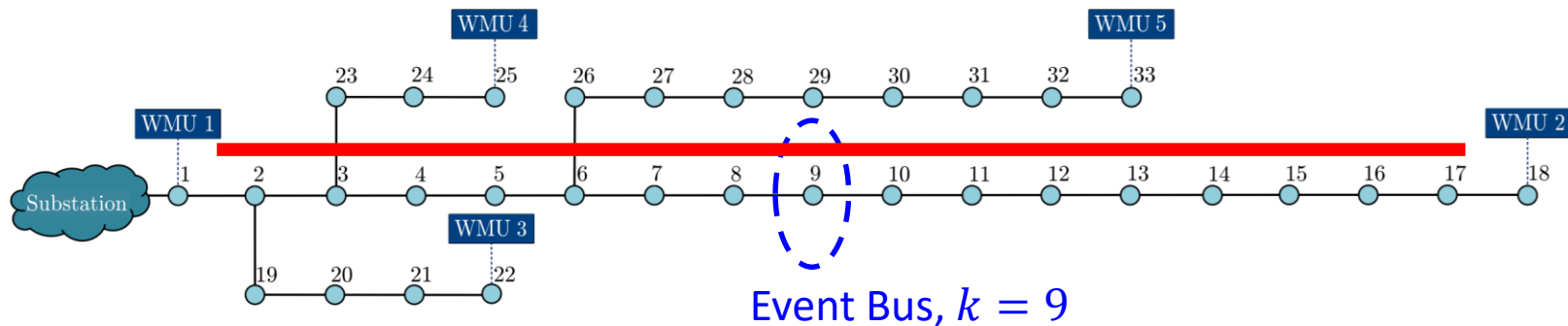
- IEEE 33-Bus Test System (PSCAD Simulations)⁵:



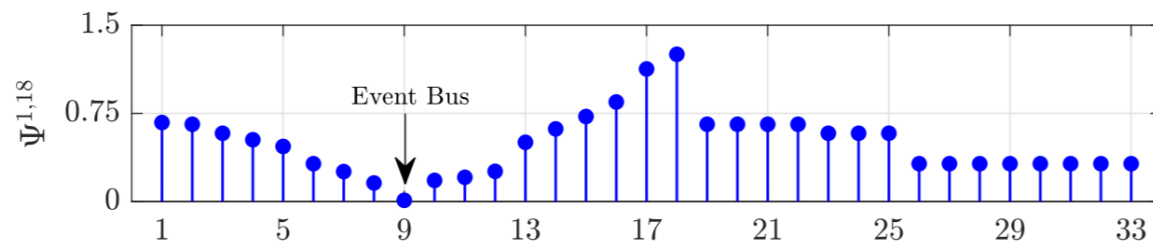
⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- IEEE 33-Bus Test System (PSCAD Simulations)⁵:



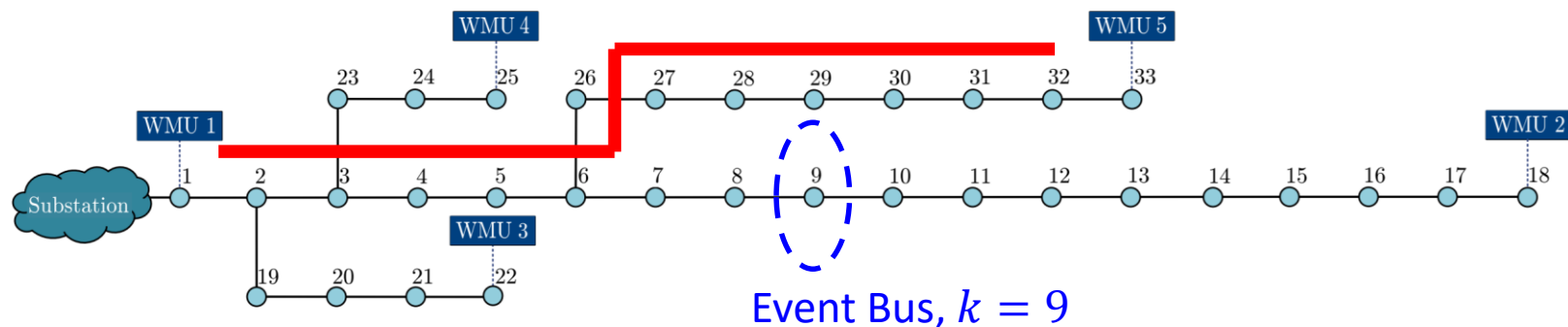
WMUs 1 and 2



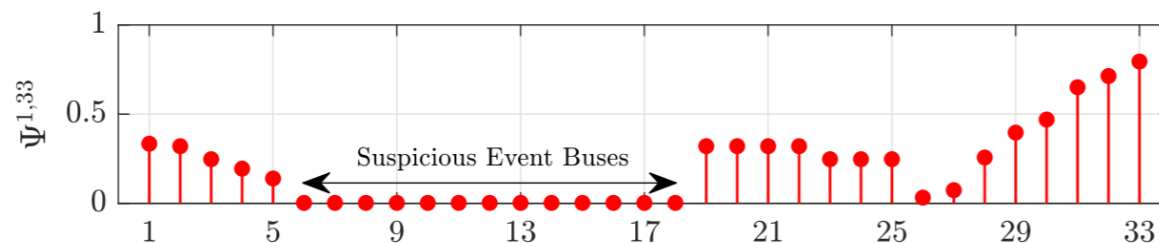
⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- IEEE 33-Bus Test System (PSCAD Simulations)⁵:



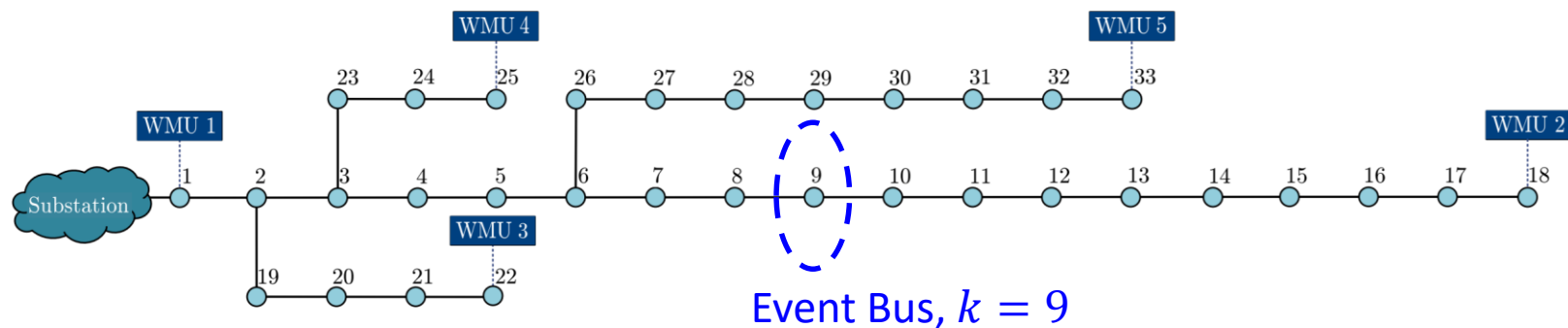
WMUs 1 and 3



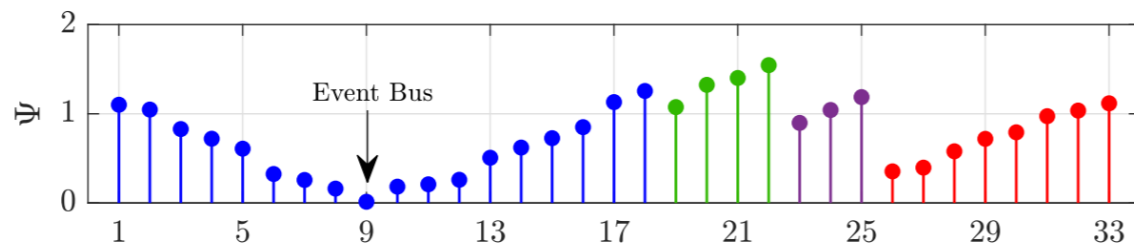
⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- IEEE 33-Bus Test System (PSCAD Simulations)⁵:



All Five WMUs



⁵ M. Izadi and H. Mohsenian-Rad, "synchronous waveform measurements to locate transient events and incipient faults in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 12, no. 5, pp. 4295-4307, Sept 2021.

Event Location Identification

- Impact of Harmonic Distortion and Measurement Noise:

	THD (%)	SNR (dB)	Correct Bus	Neighboring Bus	Other Bus
1	80		100.0 %	0.0 %	0.0 %
	50		100.0 %	0.0 %	0.0 %
	20		86.8 %	5.8 %	7.4 %
2	80		100.0 %	0.0 %	0.0 %
	50		99.9 %	0.1 %	0.0 %
	20		84.4 %	7.5 %	8.1 %
3	80		100.0 %	0.0 %	0.0 %
	50		99.8 %	0.2 %	0.0 %
	20		85.5 %	6.2 %	8.3 %

- Impact of Error in Line Parameters:

Error (%)	Correct Bus	Neighboring Bus	Other Bus
25	100.0 %	0.0 %	0.0 %
50	98.9 %	1.1 %	0.0 %
75	93.0 %	7.0 %	0.0 %
100	85.8%	14.2 %	0.0 %

Sensitivity Analysis

Synchro-Waveform Data Analysis

- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Synchro-Waveform Data Analysis

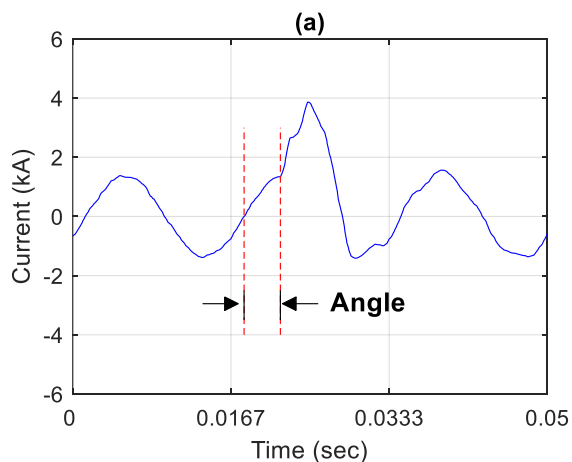
- **Situational awareness** with synchro-waveform data:
 - Data Size Per WMU: **3,981,312,000 Readings Per Day**
 - One Pair of WMUs: 8 Billion Data Points Per Day
- Event-Driven Data Analytics:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

- Example **Feature Extraction** in Waveform Measurements¹:
 - Angle, Magnitude, and Duration
 - Number of Affected Phases
 - Transient Oscillations
 - Transient Impulses
 - Fault-Specific Features
 - Changes in Steady-State Characteristics
 - Time, Season, and Location
 - Other Basic Features
 - Graphical Features

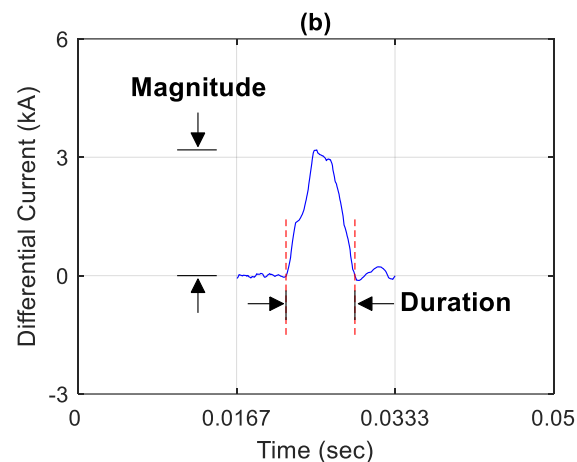
- Example **Feature Extraction** in Waveform Measurements¹:
 - **Angle, Magnitude, and Duration** ←
 - Number of Affected Phases
 - **Transient Oscillations** ←
 - **Transient Impulses** ←
 - Fault-Specific Features
 - Changes in Steady-State Characteristics
 - Time, Season, and Location
 - Other Basic Features
 - **Graphical Features** ←

Angle, Magnitude, and Duration

- These basic features can be obtained for most events.
- An example for these three features for the case of a current waveform measurement during a self-clearing fault is shown below.



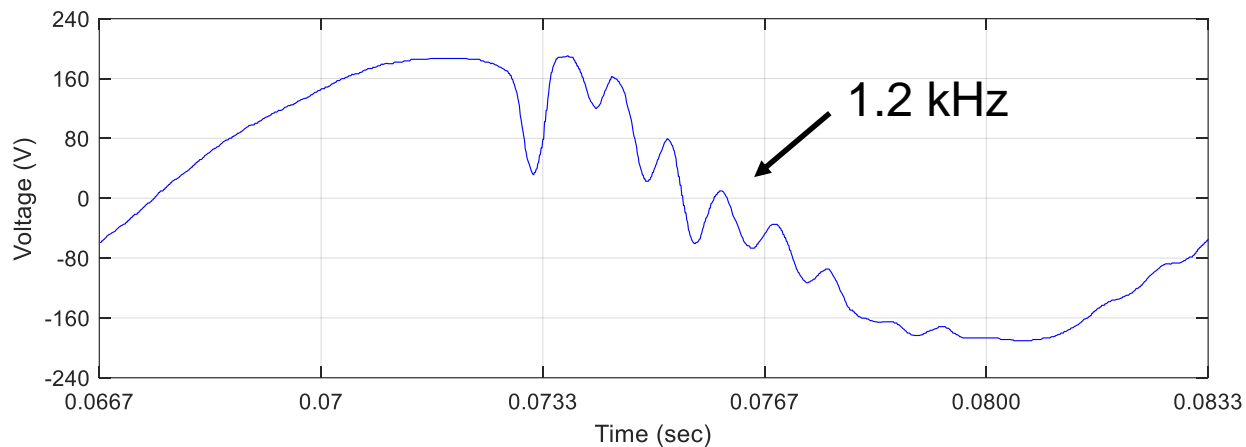
Original waveform



Differential waveform

Transient Oscillations

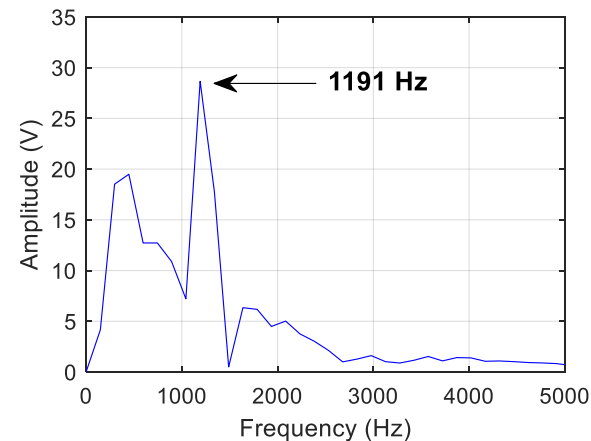
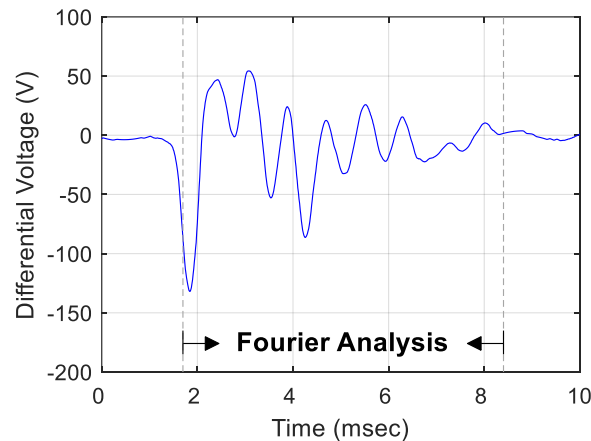
- Transient oscillations in waveform measurements are described by the *magnitude*, *duration*, and *dominant frequency* of the oscillations.



1 Cycle = 16.7 msec

Transient Oscillations

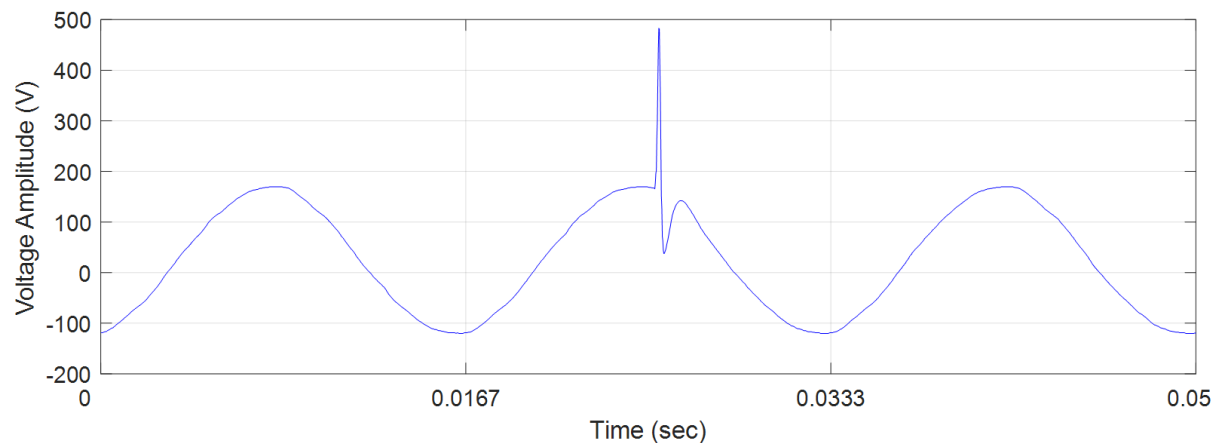
- The frequency of oscillations in waveform measurements can be obtained by using *modal analysis*; including the use of Fourier Analysis.



- The *dominant frequency* is about 1.2 KHz.

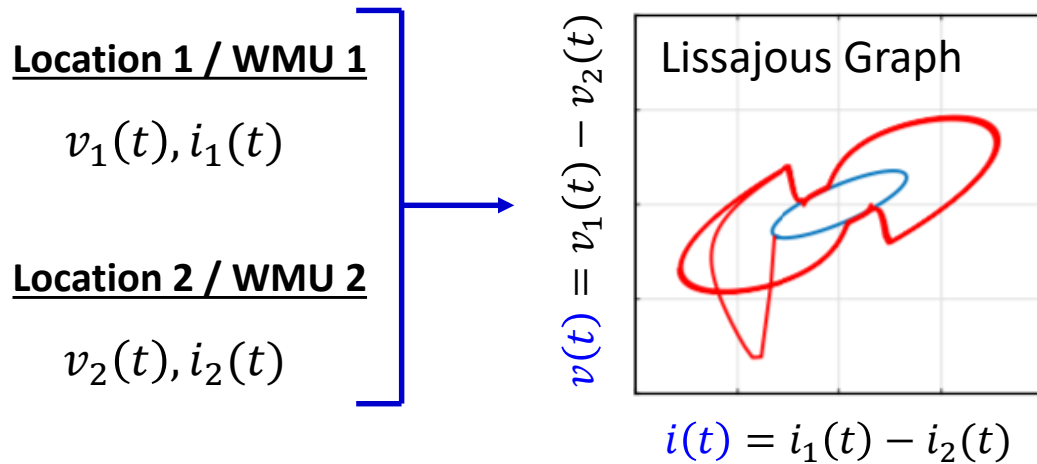
Transient Impulses

- An *impulsive transient* is a sudden change in the waveform of voltage, current, or both, that is typically unidirectional in polarity.
- A common cause of impulsive transients is *lightning strike*.



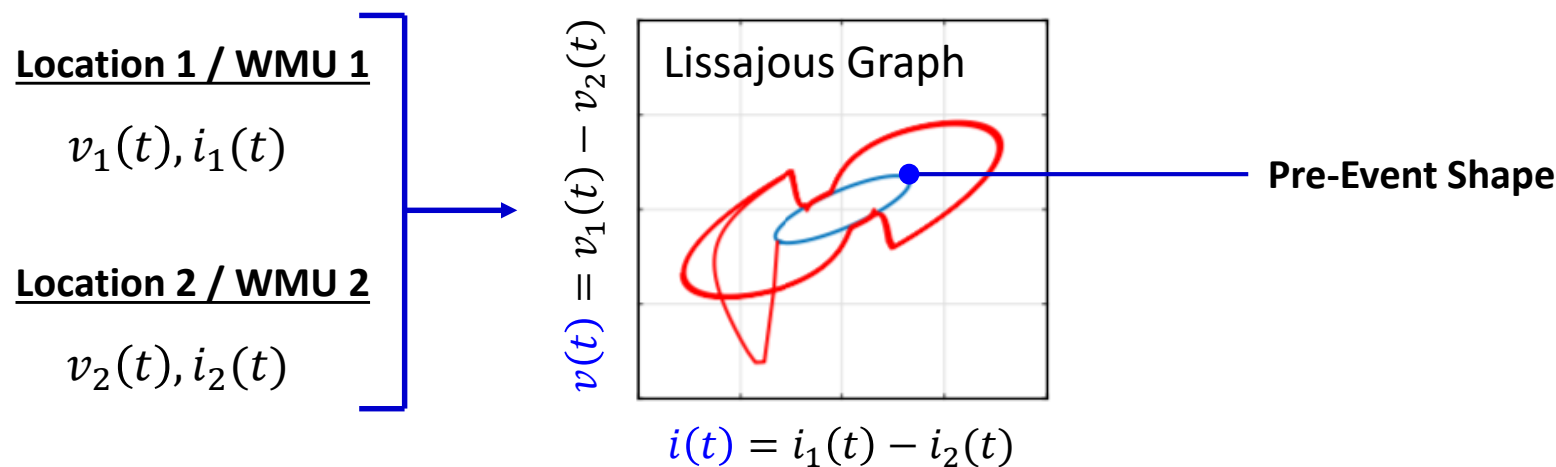
Graphical Features

- The Lissajous graphs can serve as *images* with graphical characteristics.



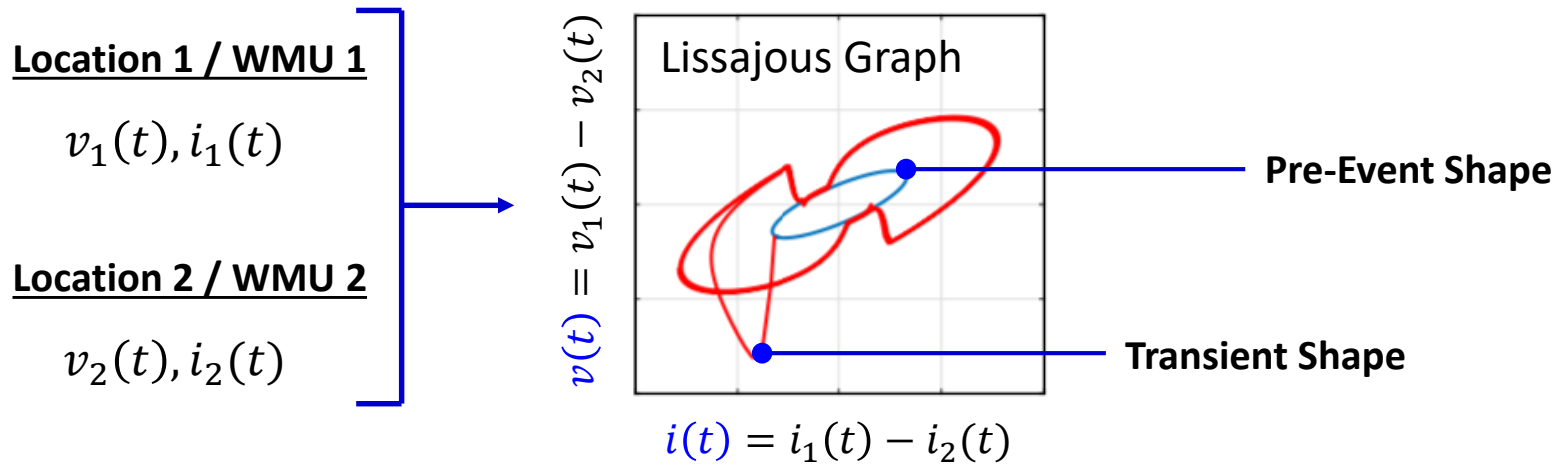
Graphical Features

- The Lissajous graphs can serve as *images* with graphical characteristics.



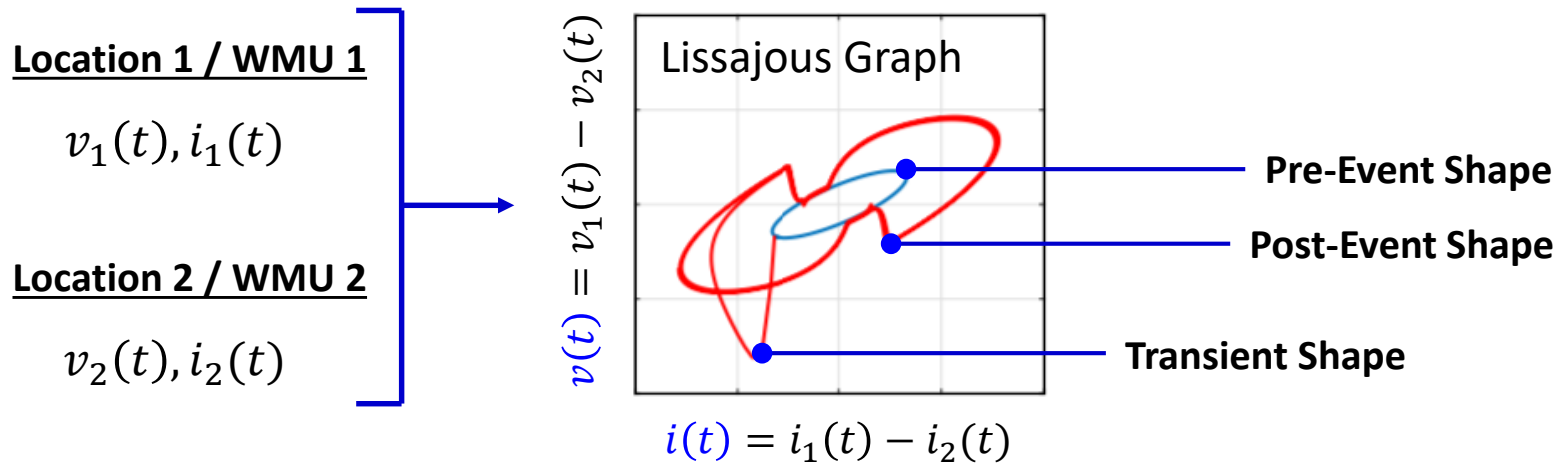
Graphical Features

- The Lissajous graphs can serve as *images* with graphical characteristics.



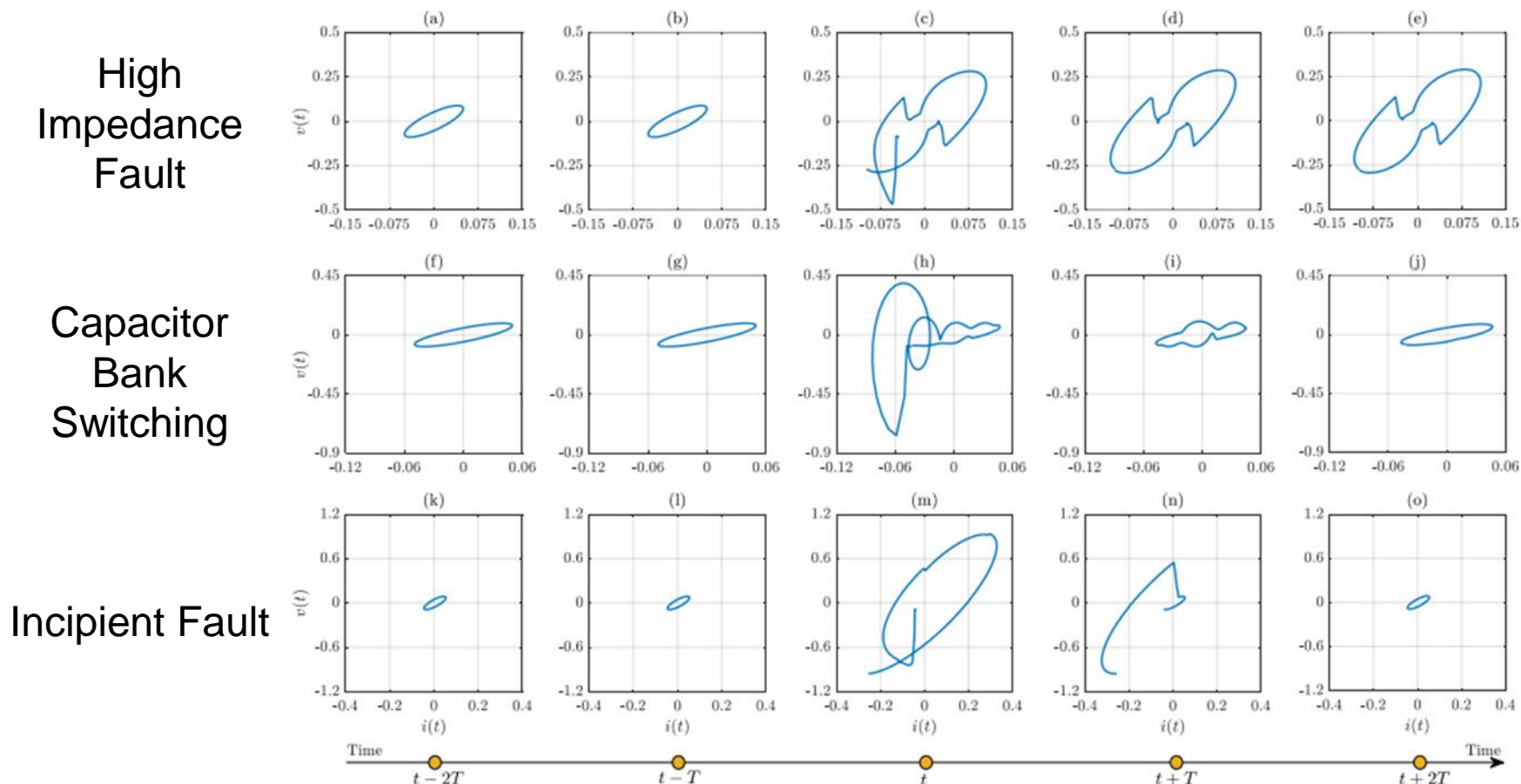
Graphical Features

- The Lissajous graphs can serve as *images* with graphical characteristics.



Event Classification

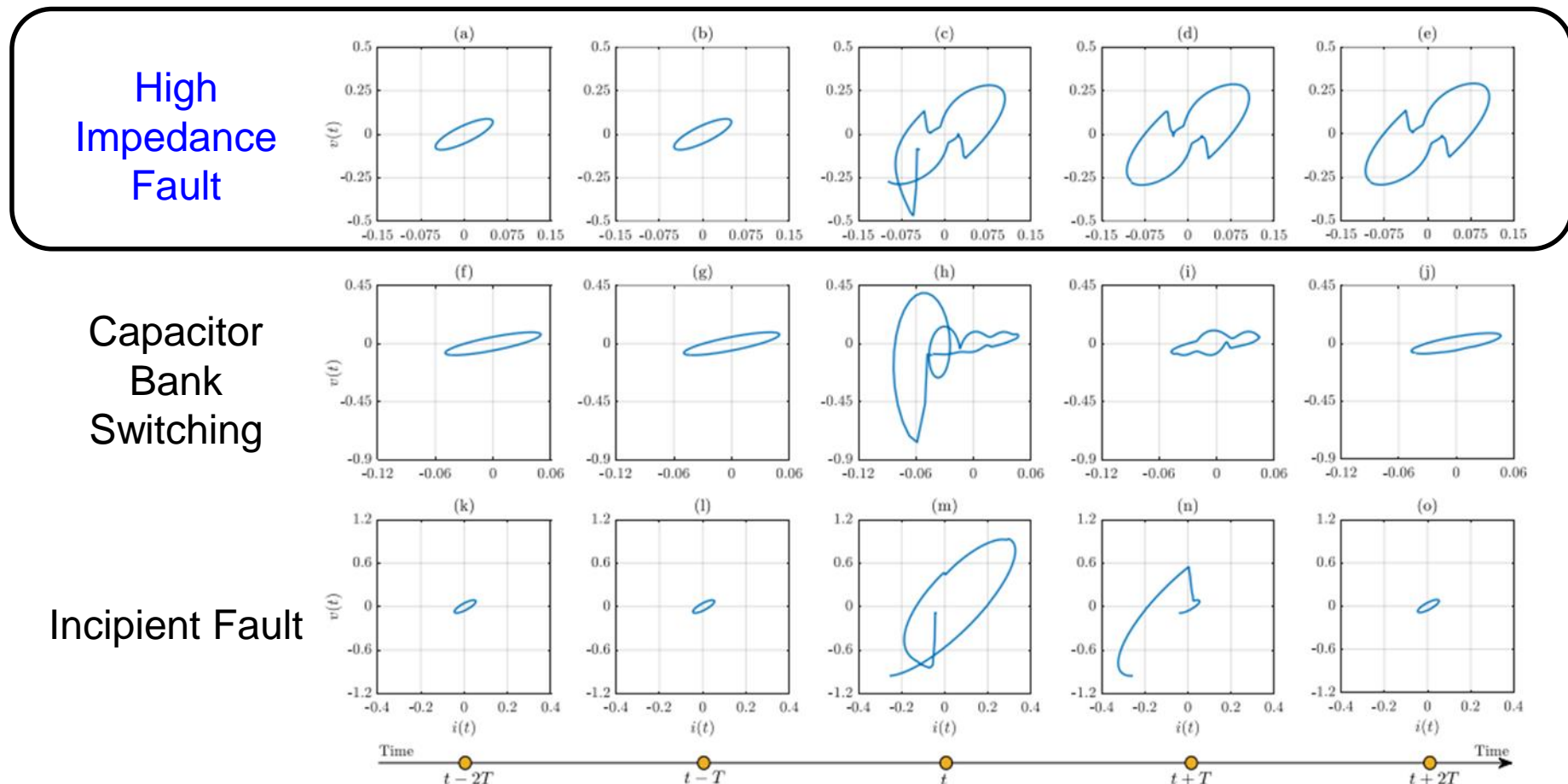
- Characterization/Classification⁶:



⁶ M. Izadi, H. Mohsenian-Rad, " Synchronized Lissajous-based method to detect & classify events in synchro-waveform measurements in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 13, no. 3, pp. 2170-2184, May 2022.

Event Classification

- Characterization/Classification⁶:

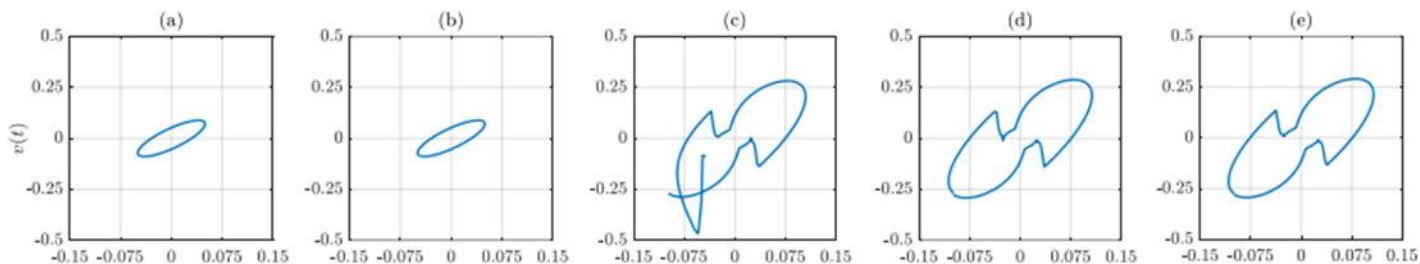


⁶ M. Izadi, H. Mohsenian-Rad, " Synchronized Lissajous-based method to detect & classify events in synchro-waveform measurements in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 13, no. 3, pp. 2170-2184, May 2022.

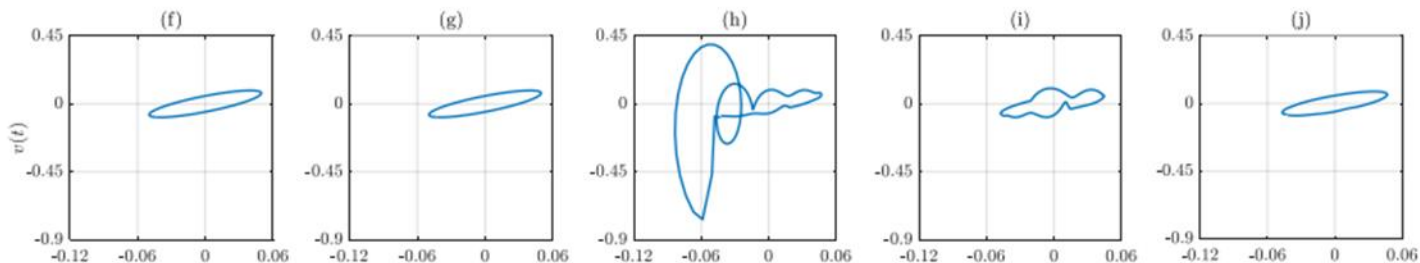
Event Classification

- Characterization/Classification⁶:

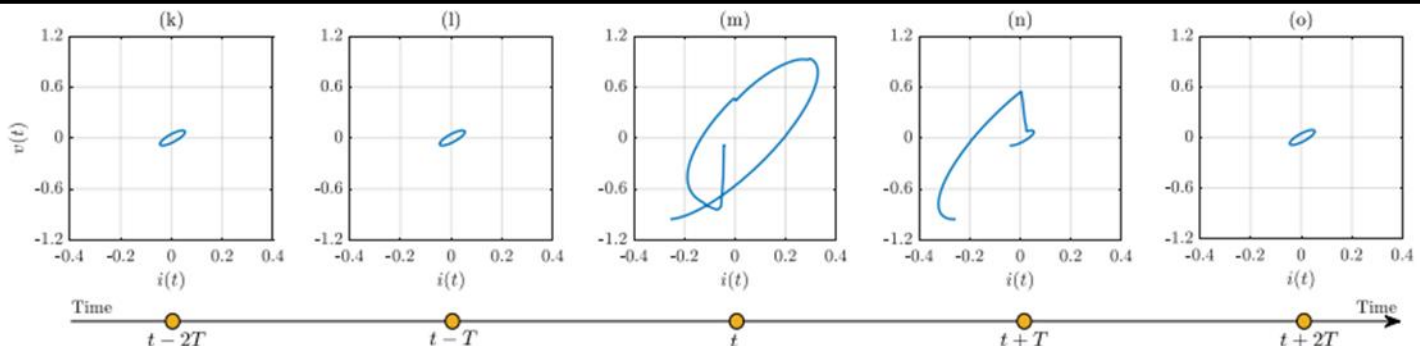
High Impedance Fault



Capacitor Bank Switching



Incipient Fault

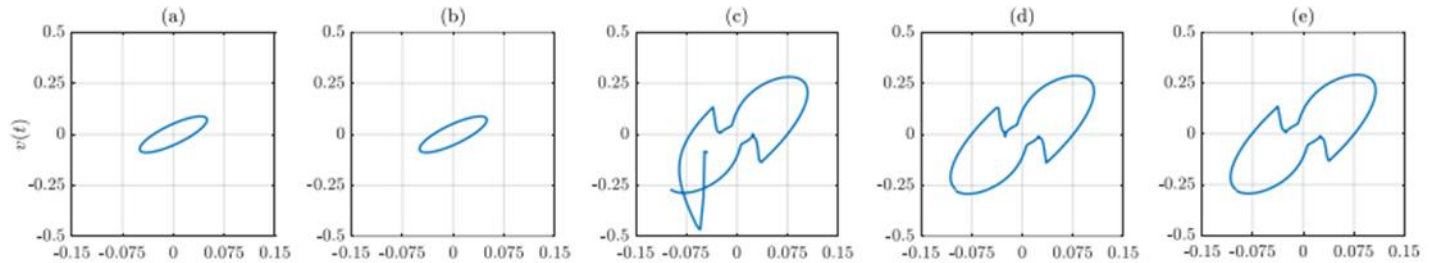


⁶ M. Izadi, H. Mohsenian-Rad, " Synchronized Lissajous-based method to detect & classify events in synchro-waveform measurements in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 13, no. 3, pp. 2170-2184, May 2022.

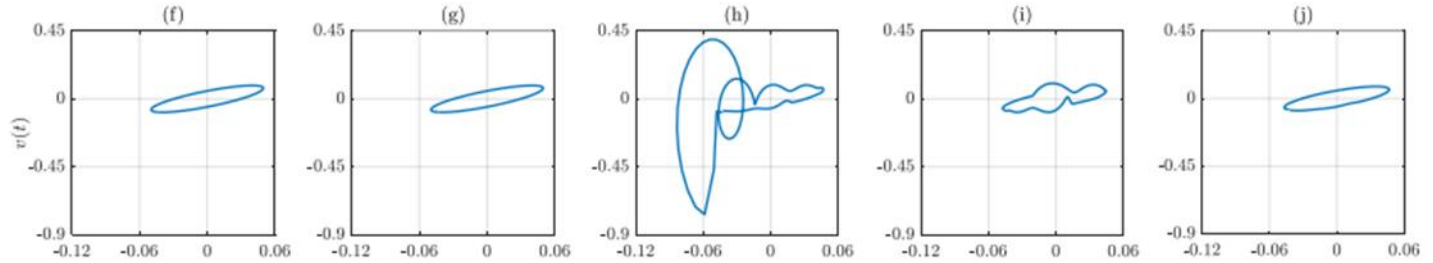
Event Classification

- Characterization/Classification⁶:

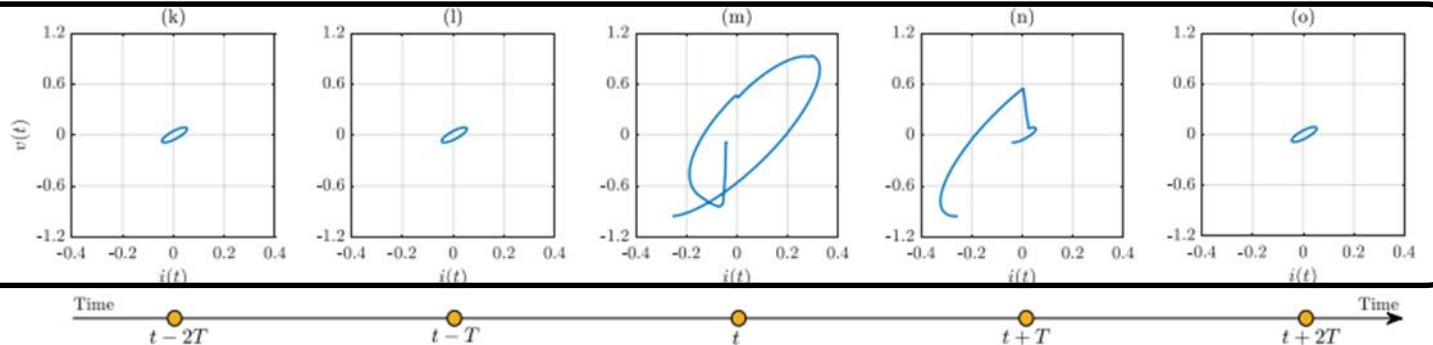
High Impedance Fault



Capacitor Bank Switching



Incipient Fault



⁶ M. Izadi, H. Mohsenian-Rad, " Synchronized Lissajous-based method to detect & classify events in synchro-waveform measurements in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 13, no. 3, pp. 2170-2184, May 2022.

Event Classification

- Classification with Convolutional Neural Network (CNN)⁶:

Confusion Matrix:

Accuracy: 97.22%

True Class	Class I	100.0%	0.0%	0.0%
	Class II	8.3%	91.7%	0.0%
	Class III	0.0%	0.0%	100.0%
		Class I	Class II	Class III

Predicted Class

Image Processing

Performance:

Class	Precision	Sensitivity	Specificity	F_1 Score
I	100.0%	92.3%	100.0%	96.0%
II	100.0%	100.0%	100.0%	100.0%
III	94.4%	100.0%	96.8%	97.1%

⁶ M. Izadi, H. Mohsenian-Rad, " Synchronized Lissajous-based method to detect & classify events in synchro-waveform measurements in power distribution networks," in *IEEE Trans. on Smart Grid*, vol. 13, no. 3, pp. 2170-2184, May 2022.

Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- **Overhead Line**
- **Underground Cable**
- **Capacitor Bank**
- **Transformer**
- **Inverters**
- **Power Electronics**
- ⋮

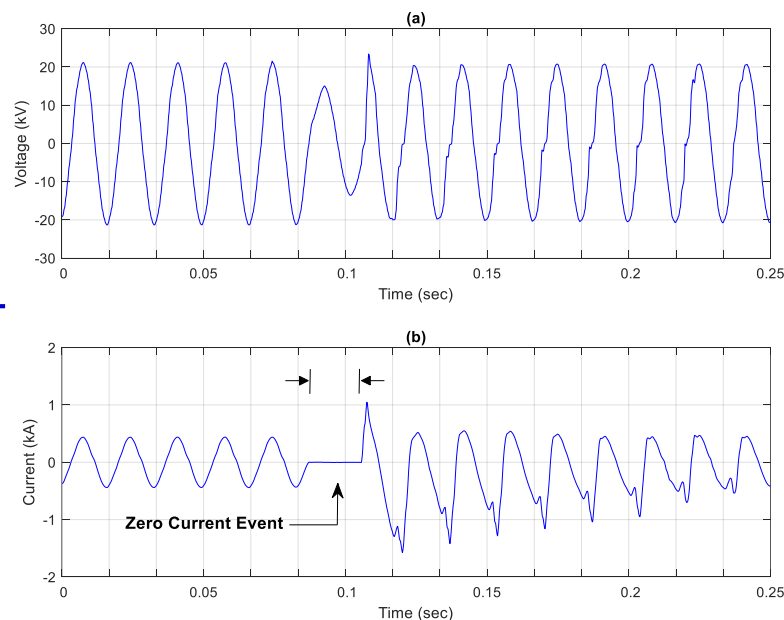
Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- Overhead Line
- Underground Cable
- Capacitor Bank
- Transformer
- Inverters
- Power Electronics
- ⋮

Unique Signatures



Lance Irwin 2010

Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- **Overhead Line** ————— **Wildfire Monitoring⁷**
- **Underground Cable**
- **Capacitor Bank**
- **Transformer**
- **Inverters**
- **Power Electronics**
- ⋮

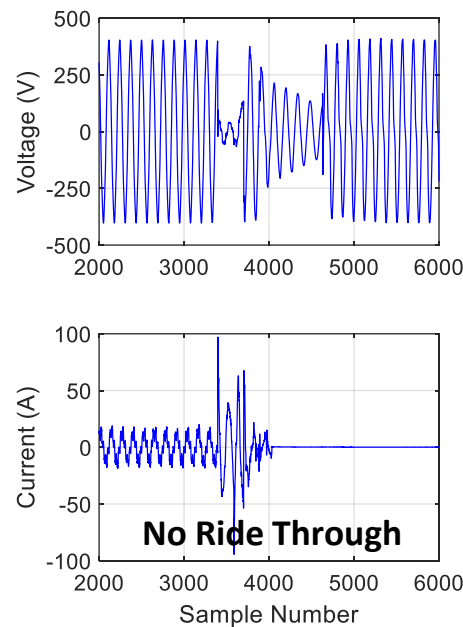
⁷ H. Mohsenian-Rad, "Synchro-Waveforms in Power Distribution with Application to Wildfire Monitoring," Panel Presentation, *IEEE PES General Meeting*, July 2021.

Synchro-Waveform Data Analysis

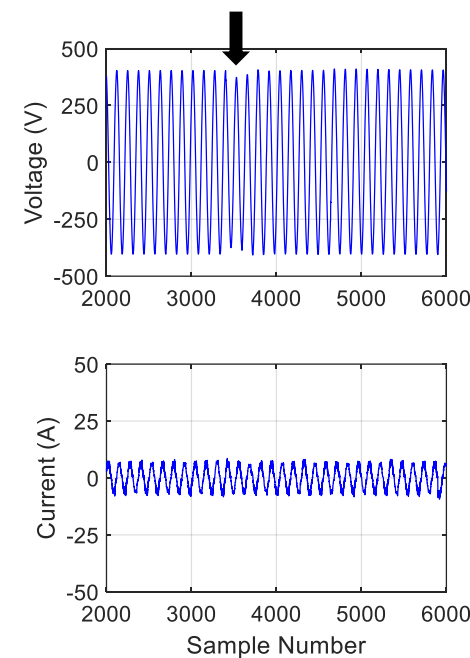
- Applications of Situational awareness with synchro-waveform data:
 - Event Detection
 - Event Location Identification
 - Event Characterization / Classification

Incipient (*Early-Stage*) Faults

- Overhead Line
- Underground Cable
- Capacitor Bank
- Transformer
- Inverters Inverter Response
- Power Electronics
- ⋮



Inverter 1



Inverter 2

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- **Overhead Line**
- **Underground Cable**
- **Capacitor Bank**
- **Transformer**
- **Inverters**
- **Power Electronics**
- ⋮

Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- Overhead Line
- Underground Cable
- Capacitor Bank
- Transformer
- Inverters
- Power Electronics
- ⋮

Analysis of Oscillations

- Oscillation Source Detection
- Sub-synchronous and Super-Synchronous
- ⋮

Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- Overhead Line
- Underground Cable
- Capacitor Bank
- Transformer
- Inverters
- Power Electronics
- ⋮

Analysis of Oscillations

- Oscillation Source Detection
- Sub-synchronous and Super-Synchronous
- ⋮

Protection Systems

- Relay Coordination
- Differential Protection
- ⋮

Synchro-Waveform Data Analysis

- Applications of Situational awareness with synchro-waveform data:
 - **Event Detection**
 - **Event Location Identification**
 - **Event Characterization / Classification**

Incipient (*Early-Stage*) Faults

- Overhead Line
- Underground Cable
- Capacitor Bank
- Transformer
- Inverters
- Power Electronics
- ⋮

And More!

Analysis of Oscillations

- Oscillation Source Detection
- Sub-synchronous and Super-Synchronous
- ⋮

Protection Systems

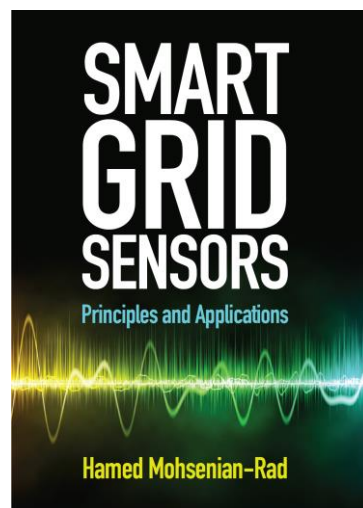
- Relay Coordination
- Differential Protection
- ⋮

- Chapter 4: Waveform and Power Quality Measurements and Their Applications

Textbook on Smart Grid Sensors:

- Working Principles
- Sample Data Sets
- Data-Driven Methods

Synchro-phasors
[Synchro-waveforms](#)
Smart meters
Building sensors
Power and energy
Probing
⋮



Cambridge University Press
April 2022
348 Pages
120 Examples
150 Exercise Questions
Solutions Manual
Instructional Slides
Data Sets

- And the references cited on the slides.

Thank You!

Hamed Mohsenian-Rad, *Ph.D., IEEE Fellow*

Professor and Bourns Family Faculty Fellow

Department of Electrical Engineering, University of California, Riverside, USA

Associate Director, Winston Chung Global Energy Center

E-mail: hamed@ece.ucr.edu

Homepage: www.ece.ucr.edu/~hamed