

Automatic Event Detection and Ring Down Analysis & Mitigation of Grid Oscillations

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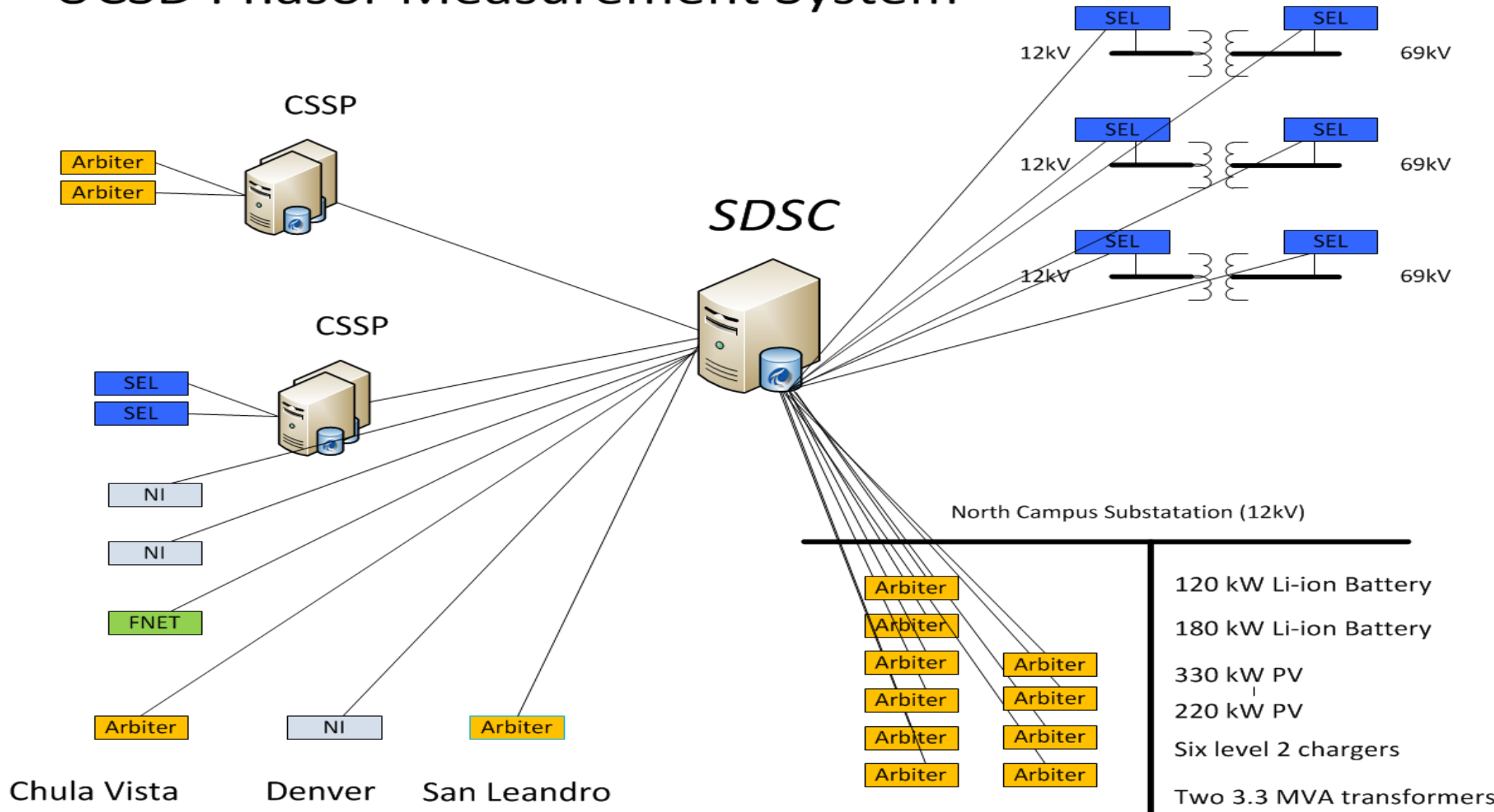


University of California, San Diego & OSISOFT

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UCSD Phasor Measurement System



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- microPMUs from PSL (ARPA-E funded)

Building Name
EBU3A Biobuilding
Atkinson Hall
Pacific Hall
Natural Sciences
CMME&CMMW
SDSC
Sverdrup
P703
Jacobs
CUP A
CUP B
North Campus Housing
Rady School
RIMAC
Hospital CC Embergency A
Hospital CC Emergency B
SOM Pharm
SOM BSB
SIO Hubbs Hall
CPS WC -9

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Identification/classification:

- Identify major modes of grid oscillation
- Identify their frequencies, damping and modal participation
- Develop dynamic model that can be used for future control/mitigation of disturbances

Analysis/control:

- Determine how well models correlate with modes
- Use models for automatic control for mitigation

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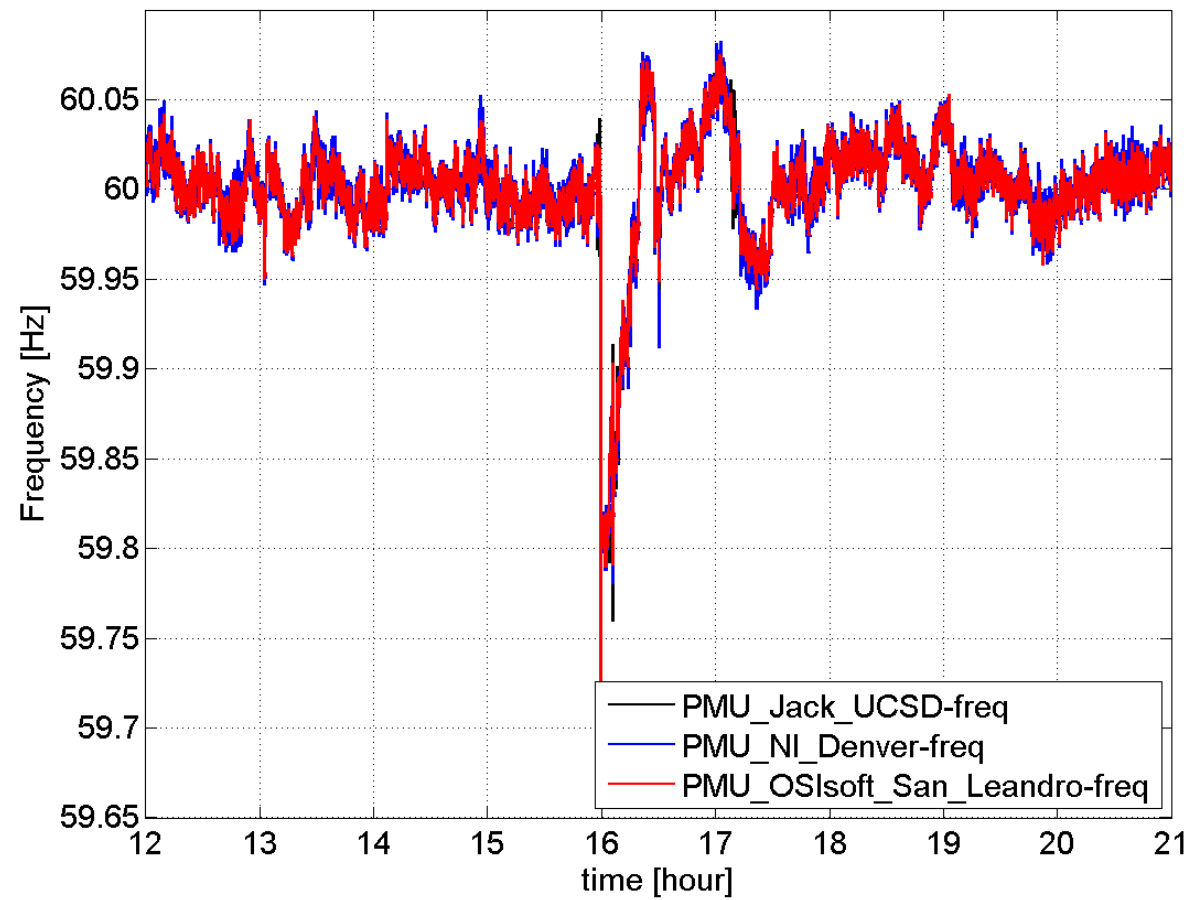
- Detection of Events via **Filtered Rate of Change (FRoC) signal**
 - Auto Regressive Moving Average (ARMA) filter of **ambient data**.
 - Definition of Filtered Rate-of-Change (Froc) signal for Event Detection
- Ring Down Analysis of Events via **Realization Algorithm**
 - Discrete-Time State Space Modeling of **disturbance data**.
 - Modeling of grid real power dynamics
- Mitigation of Events via Real-time Control
 - Use **dynamic model** from Realization Algorithm
 - Design low-order real-time (automatic) control with minimal control effort
- Illustration in this talk:
 - Part 1: **Automatic event detection applied to May 30 WECC event**
 - Part 2: **UCSD Microgrid: analysis and control of Oct. 9 event**

PART 1

Automatic Event Detection Application to May 30 WECC disturbance

Grid events/oscillations (example: May 30 WECC event)

- PMU generated frequency signal
- How do we detect individual events?
- How can we quantify these events?
- What do these events tell us about our (micro)grid?

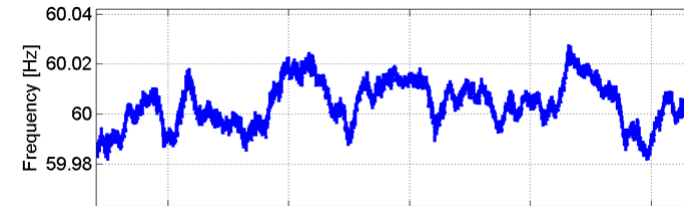


May 30 data: 972000 data points (30Hz sampling noon-9pm)

- In ambient situation we may assume:
 - Fluctuations in frequency signal $F(k)$ assumed due to “random noise” on grid
 - $F(k)$ can be modeled as a “filtered white noise”

$$F(k) = H(q)e(k)$$

where $H(q)$ is an unknown filter and $e(k)$ is a white noise.



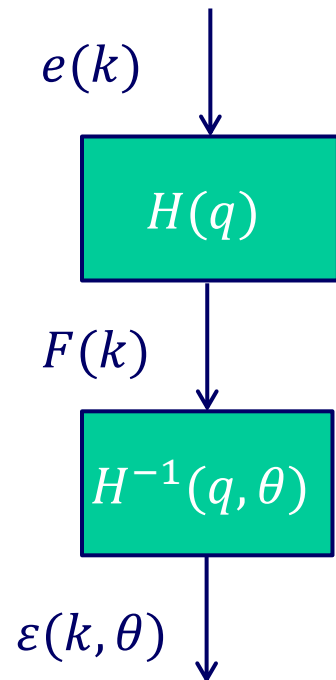
- Possible approximation for filter $H(q)$: ARMA filter

$$H(q, \theta) = \frac{b_0 + b_1q^{-1} + \dots + b_nq^{-n}}{1 + a_1q^{-1} + \dots + a_nq^{-n}}$$

- Filter $H(q)$ is stable and stably invertible
- We can compute

$$\varepsilon(k, \theta) = H(q, \theta)^{-1}F(k)$$

- Parameters $\theta = [b_1 \ \dots \ b_n \ a_1 \ \dots \ a_n]$ can be estimated via **Least Squares (Prediction Error)** to minimize variance of error $\varepsilon(k, \theta)$.



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- With optimal value of θ we have “smallest possible”

$$\varepsilon(k, \theta) = H(q, \theta)^{-1} F(k)$$

during ambient behavior.

- To create FRoC: add additional filtering on $\varepsilon(k, \theta)$ to monitor Rate of Change in $F(k)$
- Typical Filter:

$$FRoC(k) = R(q)L(q)H(q, \theta)^{-1}F(k)$$

$$R(q) = \frac{q - 1}{q - 0.9}, \quad L(q) = \frac{0.1367q + 0.1367}{q - 0.7265}$$

END RESULT: a **real-time recursive formula** to compute $FRoC(k)$:

$$FRoC(k) = b_0F(k) + b_1F(k - 1) + \dots + b_nF(k - n) - a_1FRoC(k - 1) - \dots - a_nFRoC(k - n)$$

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- In our case based on real-time PMU data we created the discrete-time filter equation to obtain FRoC(k):

$$\begin{aligned}FRoC(k) = & 0.12786 \cdot F(k) - 0.25412 \cdot F(k - 1) - 0.00094 \cdot F(k - 2) \\ & + 0.25411 \cdot F(k - 3) - 0.12694 \cdot F(k - 4) \\ & + 3.48506 \cdot FRoC(k - 1) - 4.54036 \cdot FRoC(k - 2) \\ & + 2.61982 \cdot FRoC(k - 3) - 0.56464 \cdot FRoC(k - 4)\end{aligned}$$

- Compared with ROCOF(k):

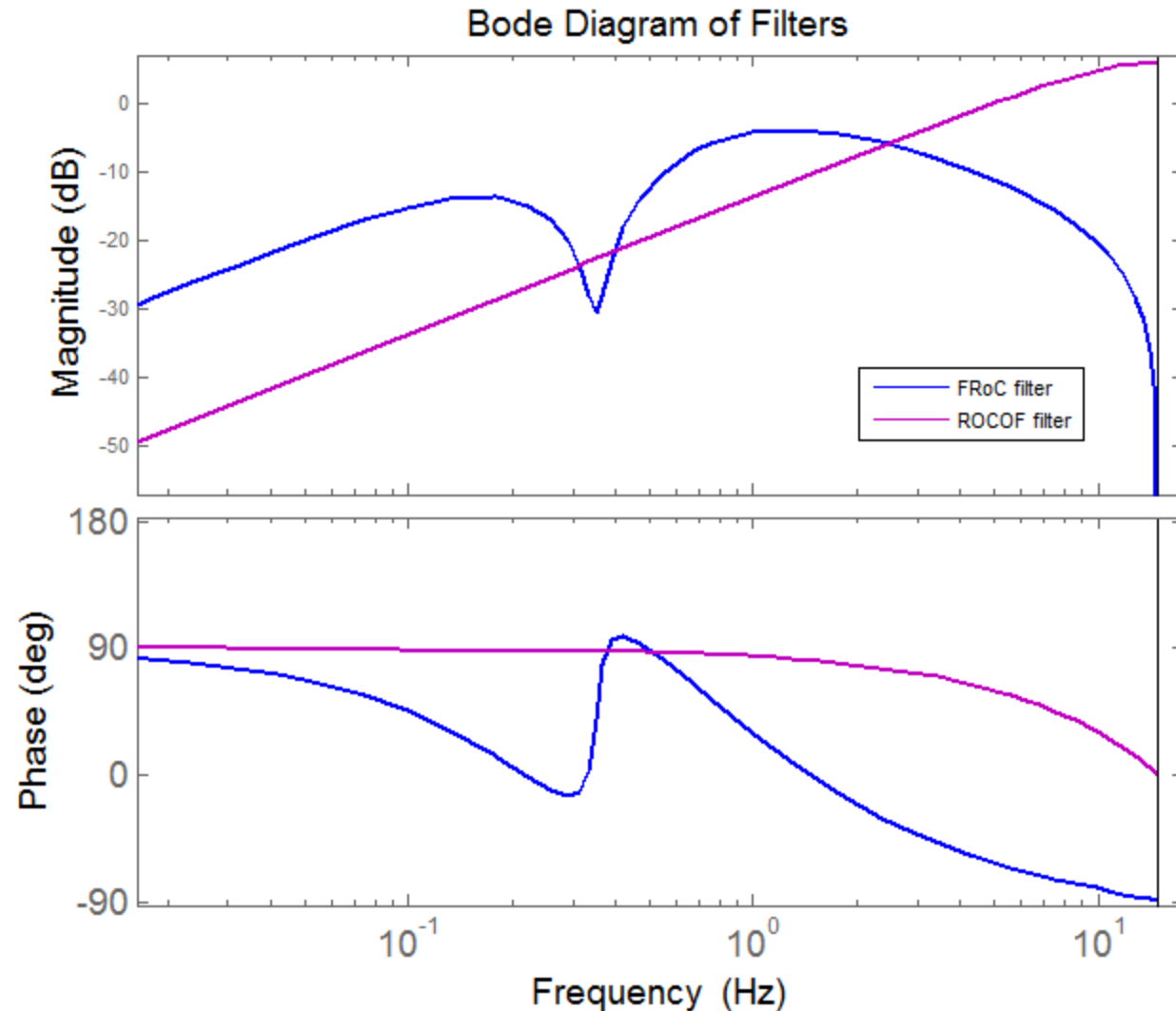
$$ROCOF(k) = 30(F(k) - F(k - 1))$$

(dirty discrete-time derivative)

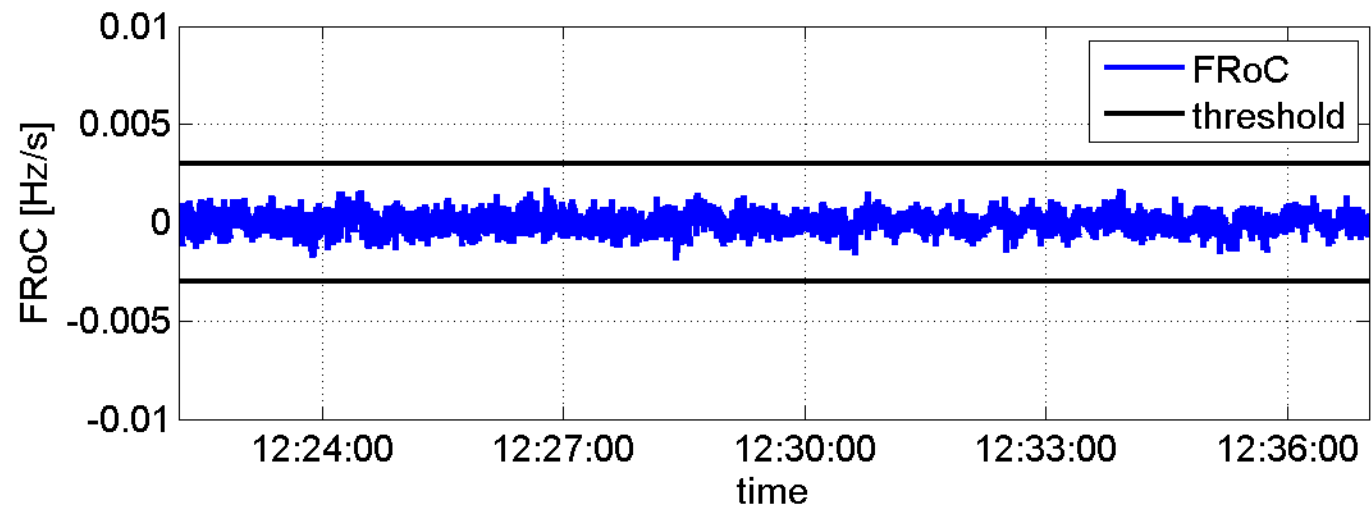
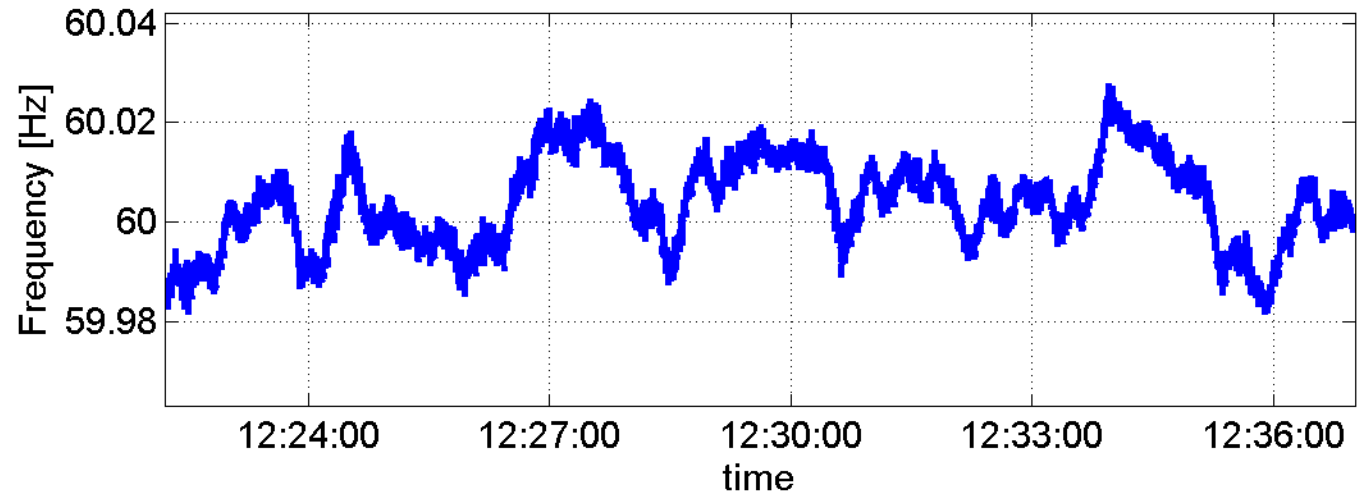
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- Bode plot of filters used to create FRoC(k) and ROCOF(k) illustrates the benefits:

- Filter looks like a 'differentiator'
- Additional filtering of harmonic disturbances ambient data at 0.35Hz
- Additional low pass filter to reduce noise



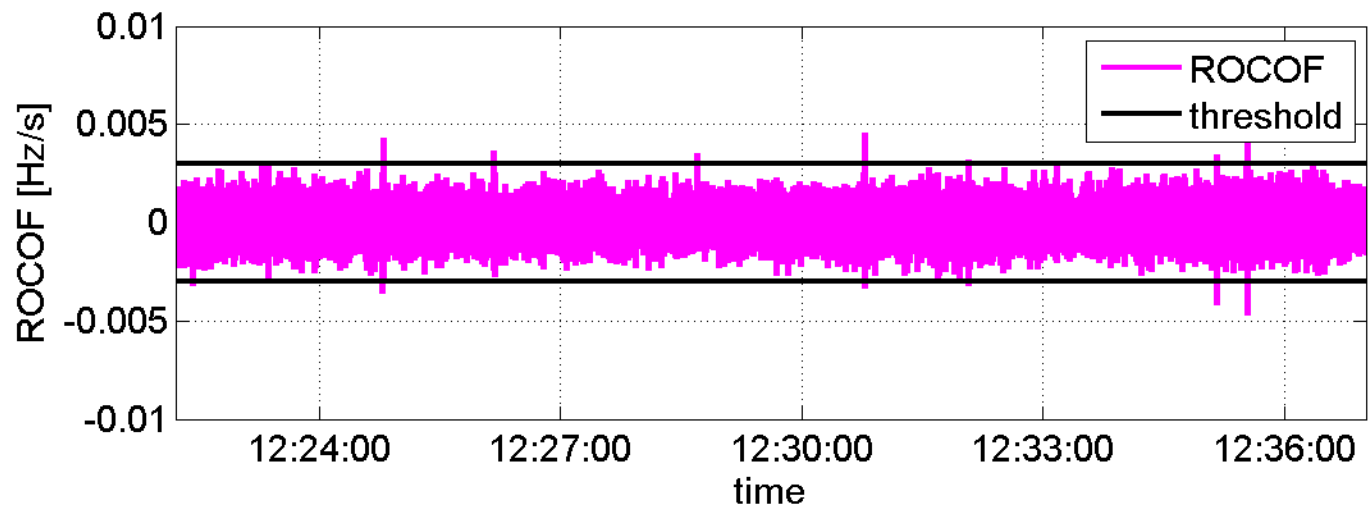
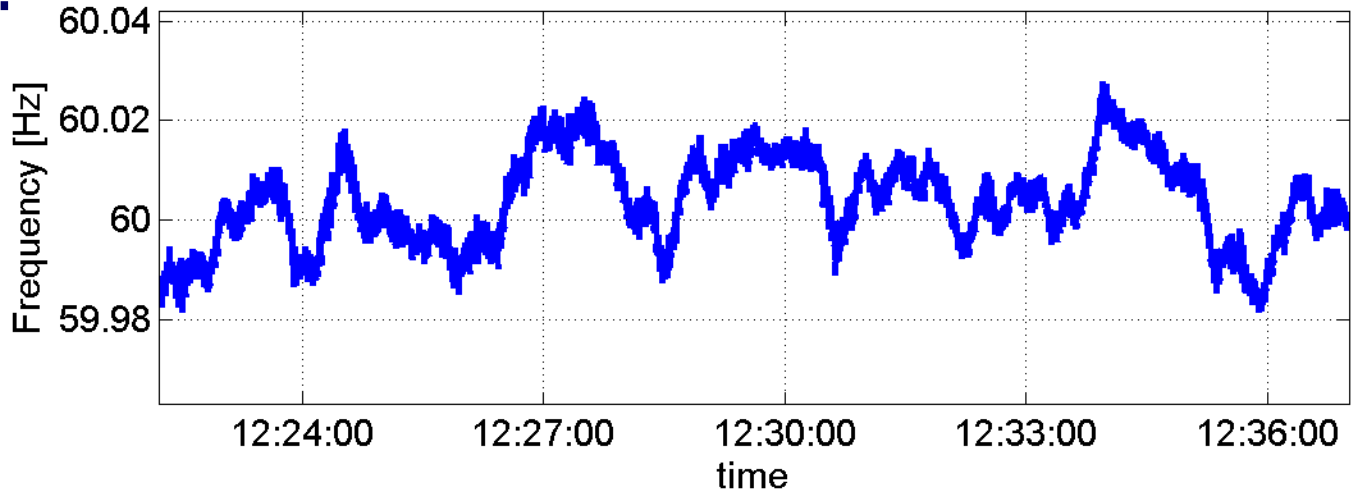
- Small $F_{RoC}(k)$ during ambient behavior
- Even for “noisy” NI PMU



Compare with ROCOF:

$$ROCOF(k) = \frac{F(k) - F(k - 1)}{\Delta t}$$

(dirty discrete-time derivative)



- Much larger than $FROc(k)$
- Would require larger thresholds

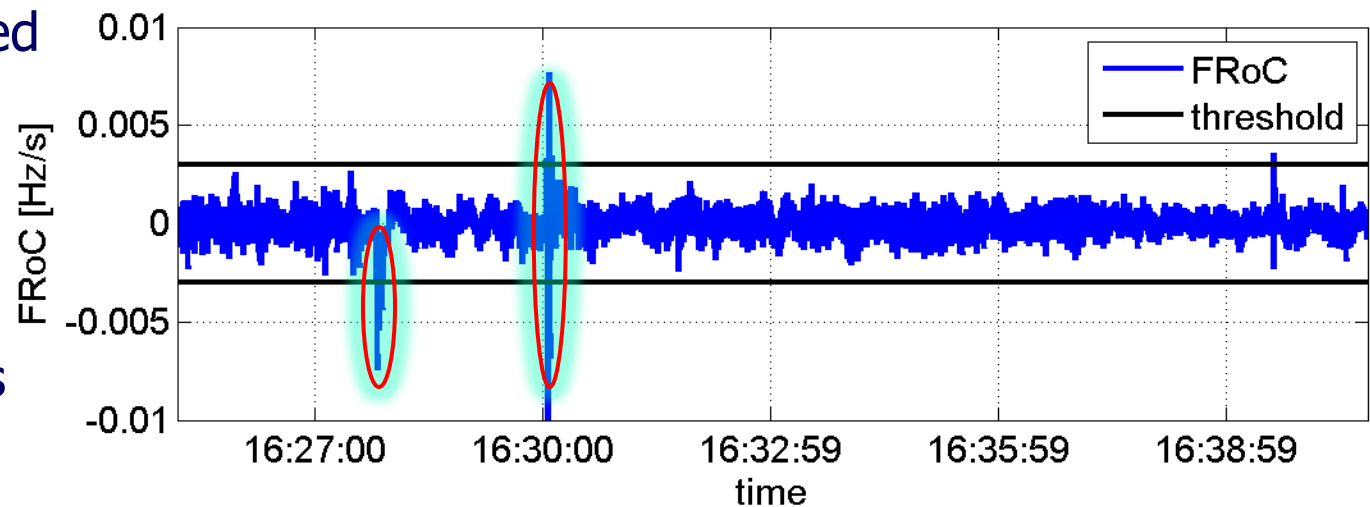
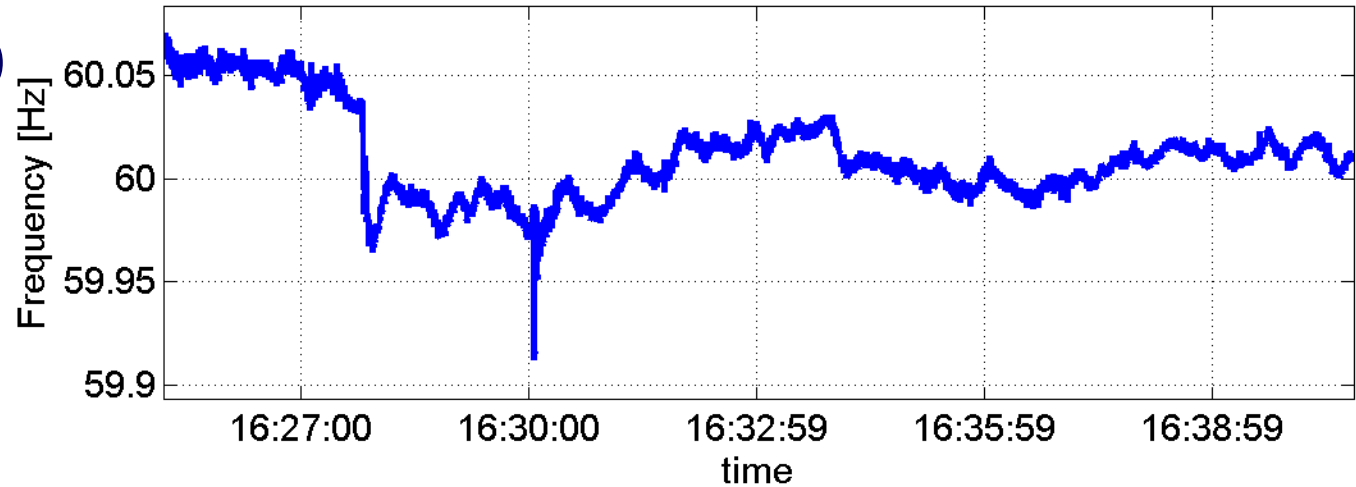
- Small thresholds with small $F_{RoC}(k)$ during ambient behavior

- Detection of events via:

- Set **threshold** based on ambient data

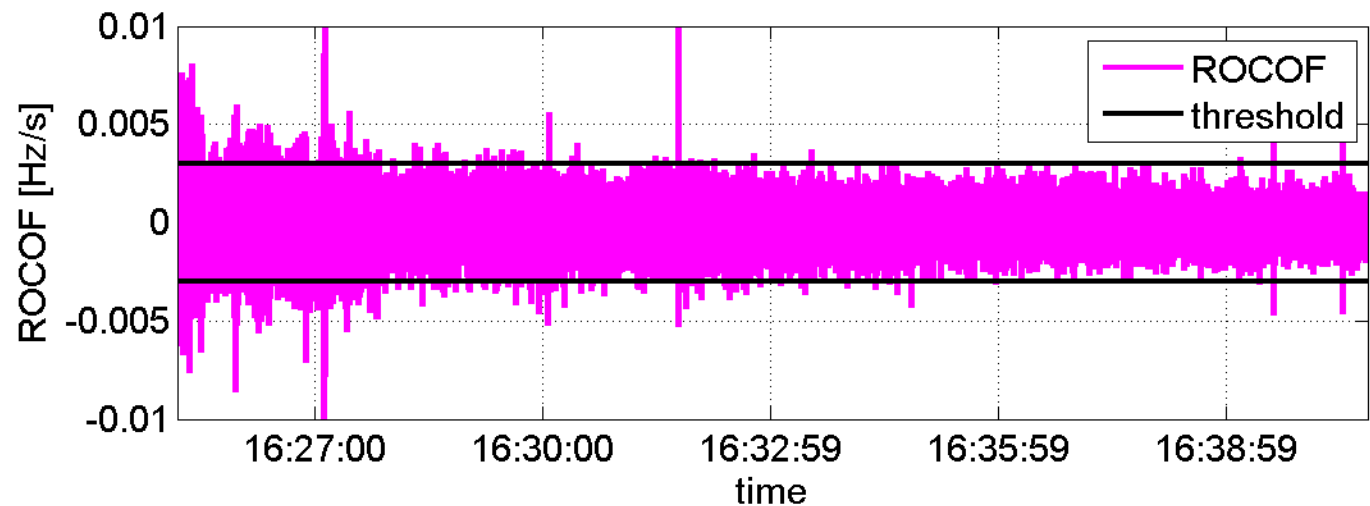
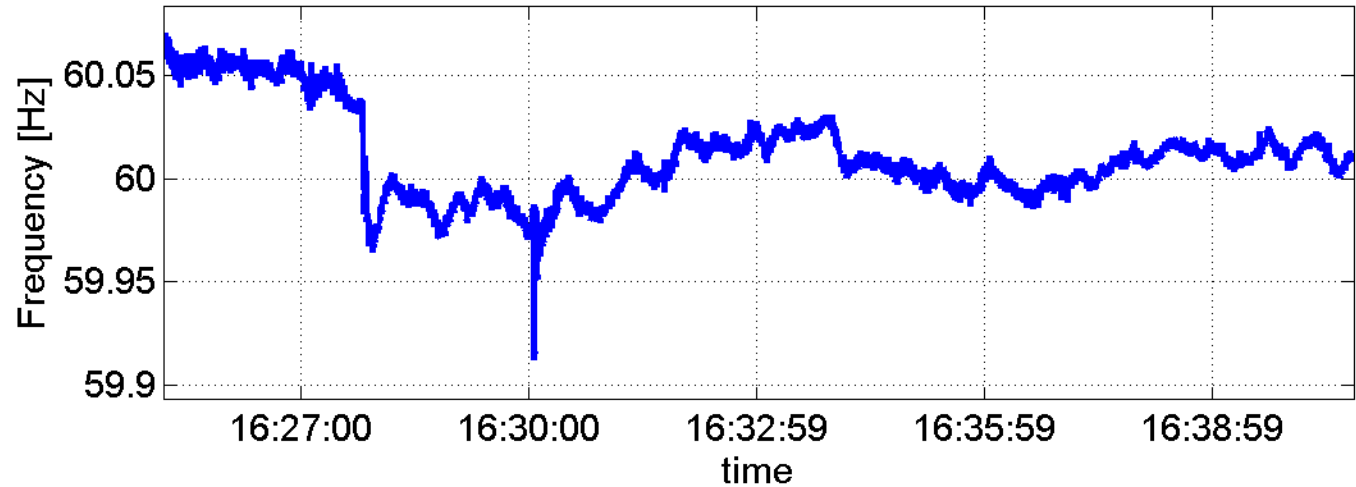
- $F_{RoC}(k)$ **outside threshold for m consecutive points**

- Classify event by saving/analyzing N data points



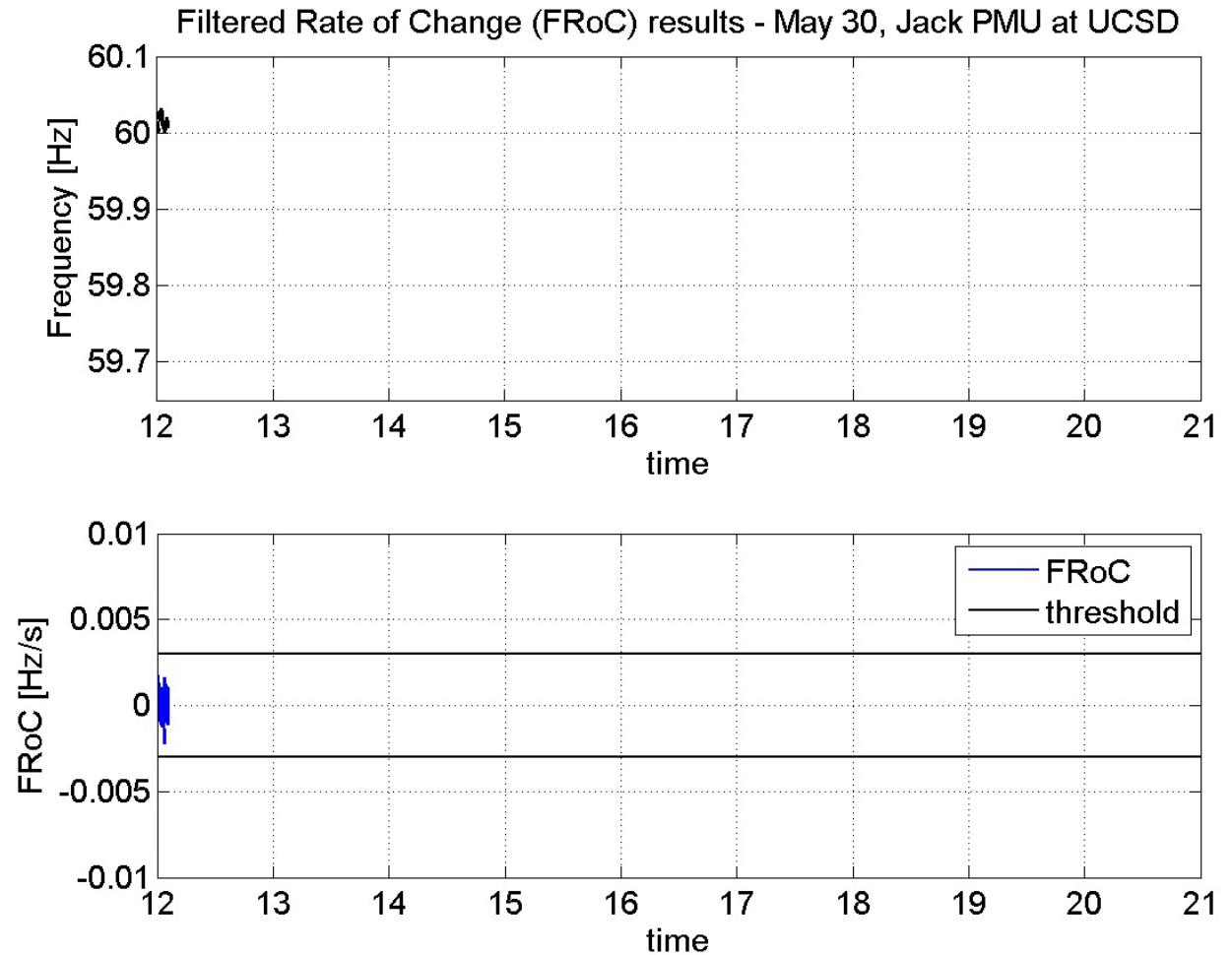
Compare with ROCOF

- Much larger than $FROC(k)$
- More false alarms



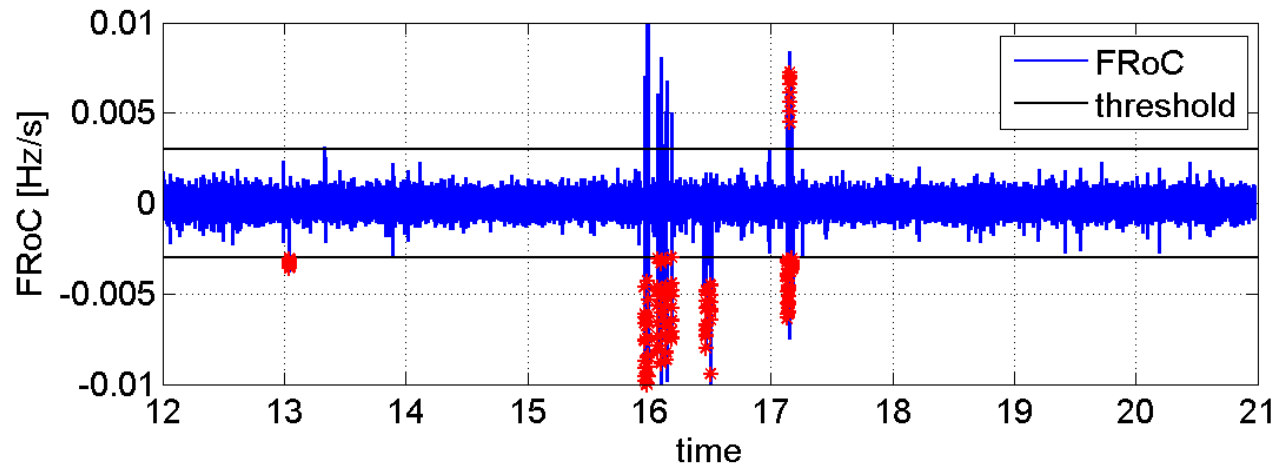
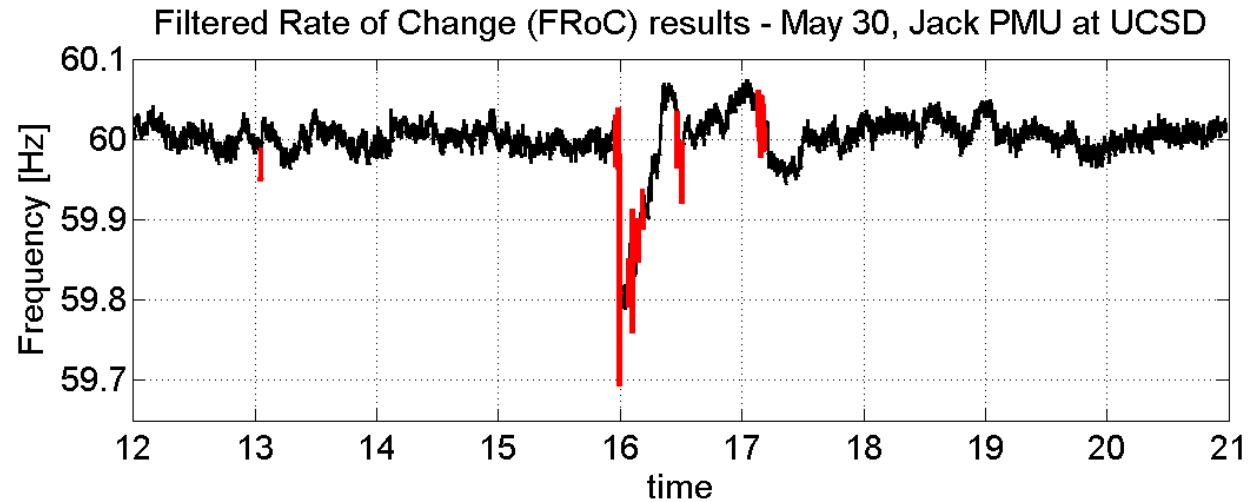
Automatically:

- Detect event.
(via threshold on Filtered Rate of Change signal)

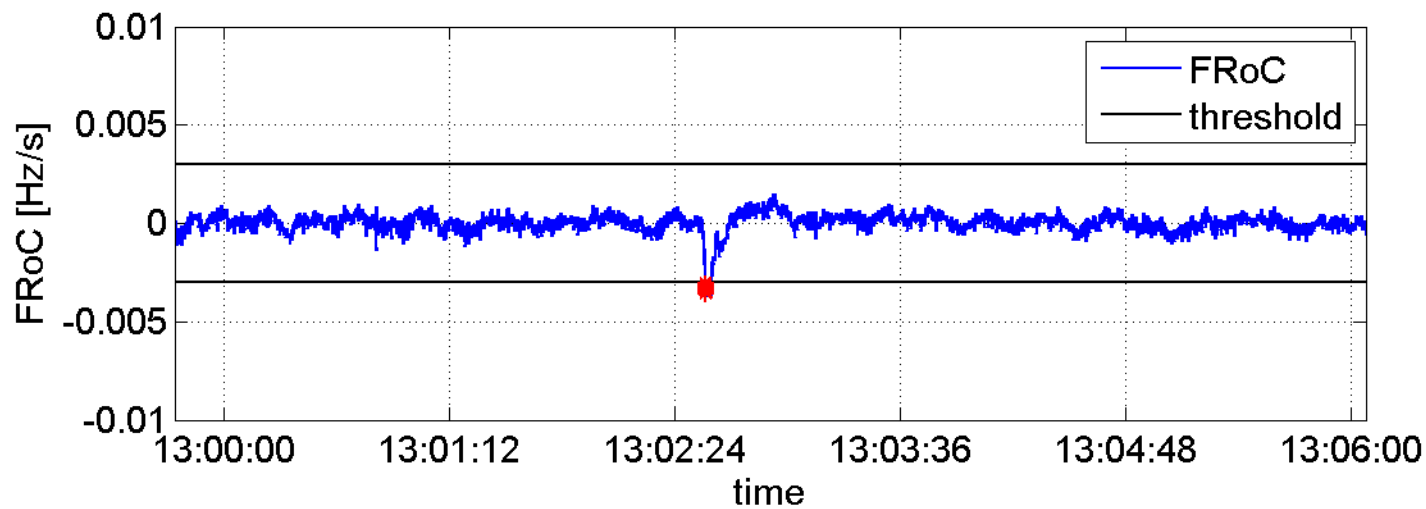
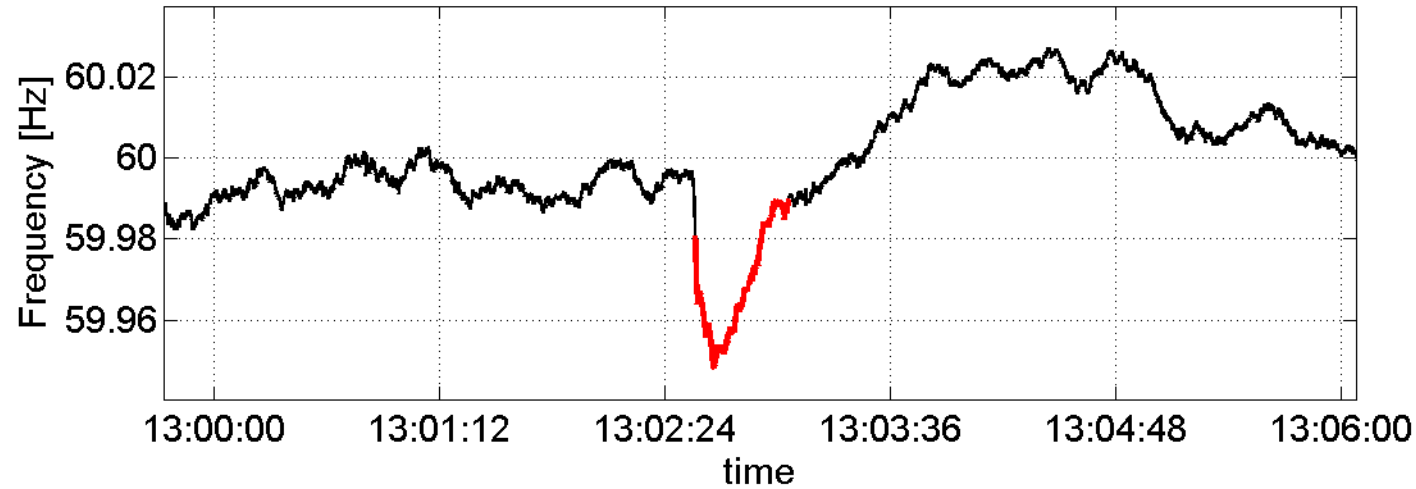


Automatically:

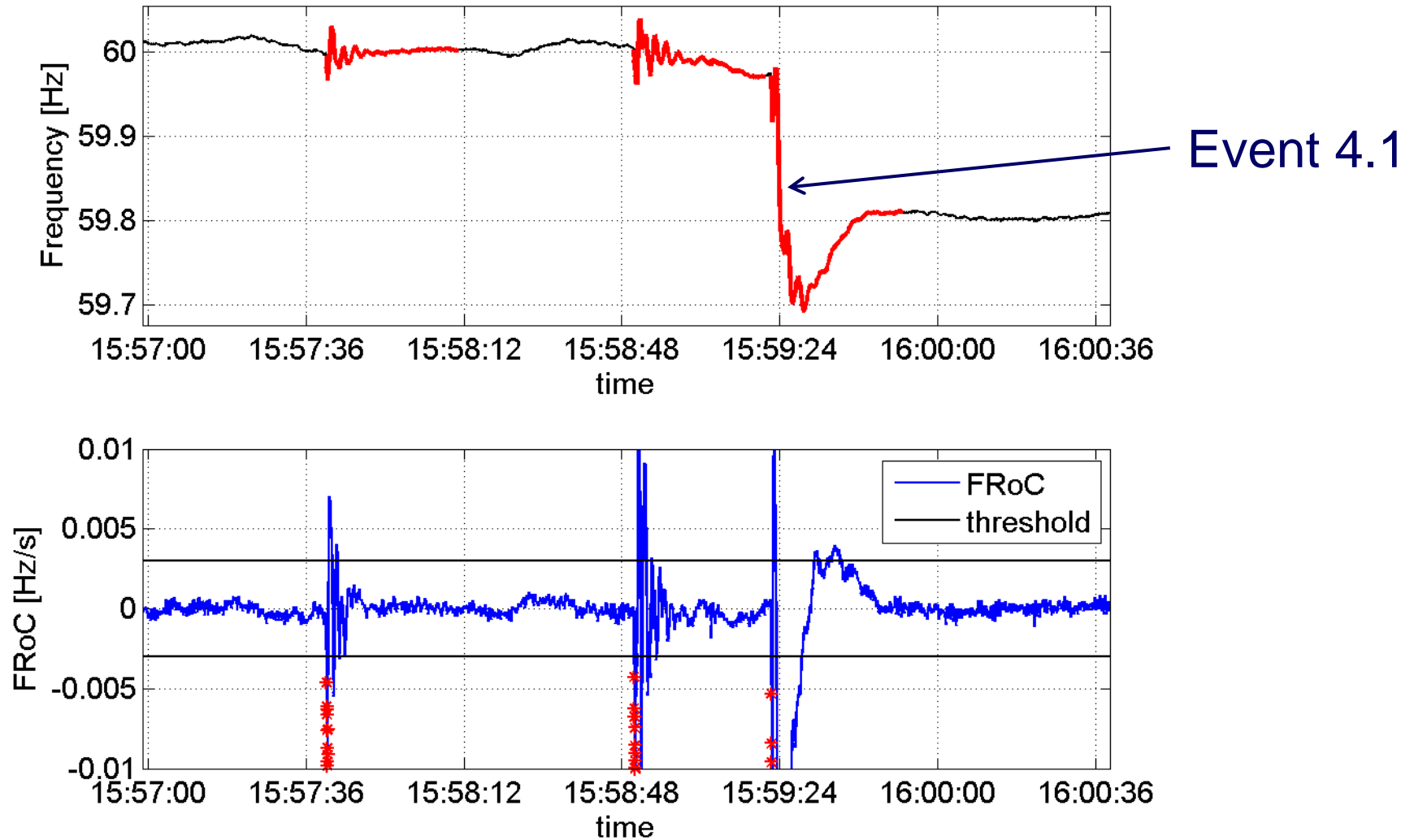
- Detect event.
(via threshold on Filtered Rate of Change signal)
- Able to distinguish 14 separate events over 9 hour data



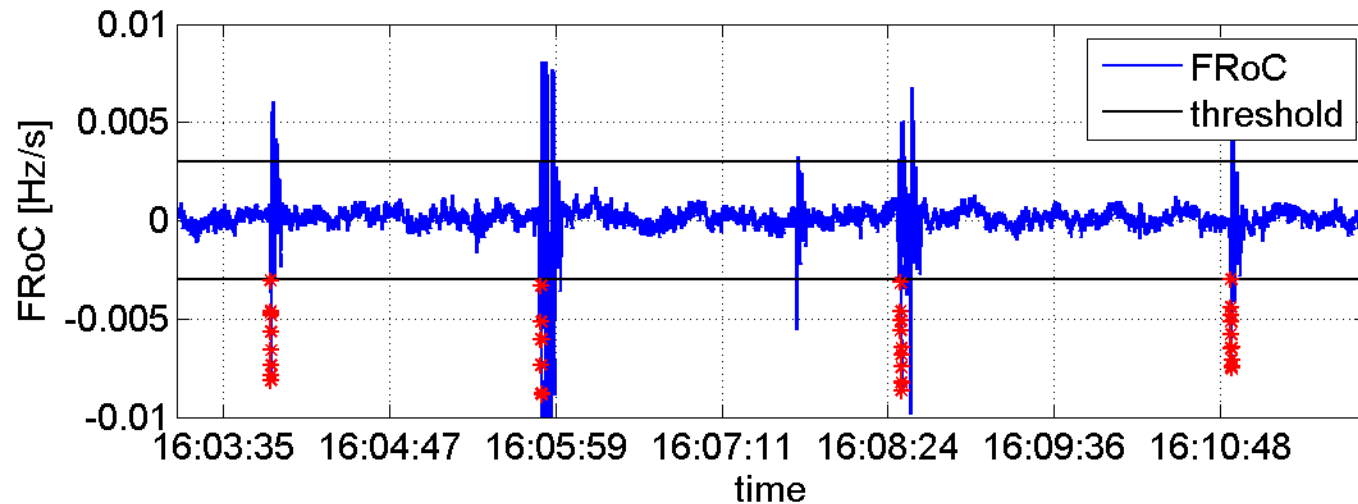
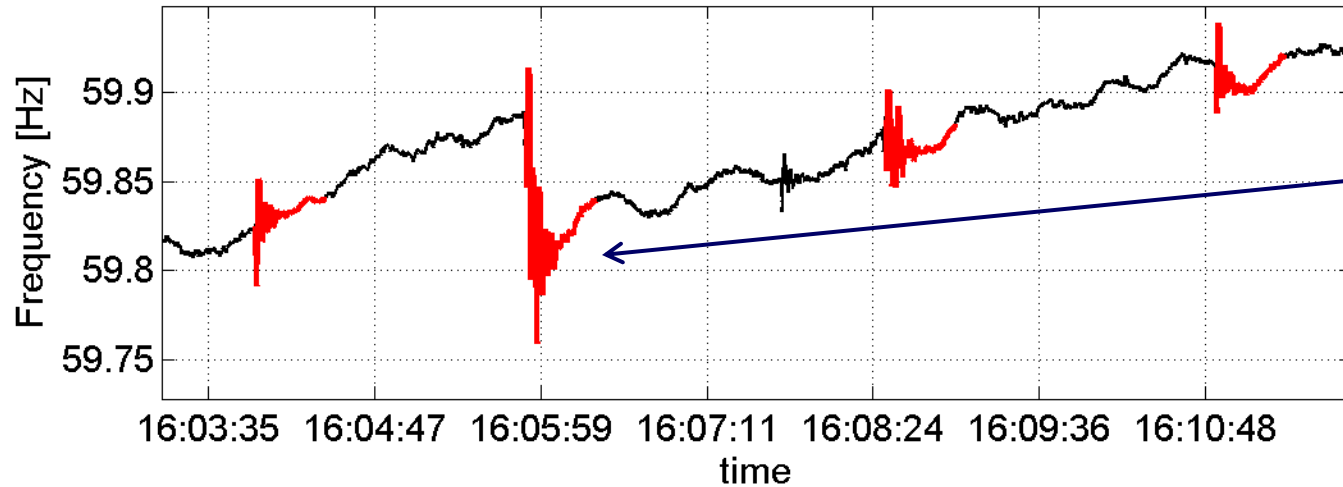
Filtered Rate of Change (FRoC) results - May 30, Jack PMU at UCSD



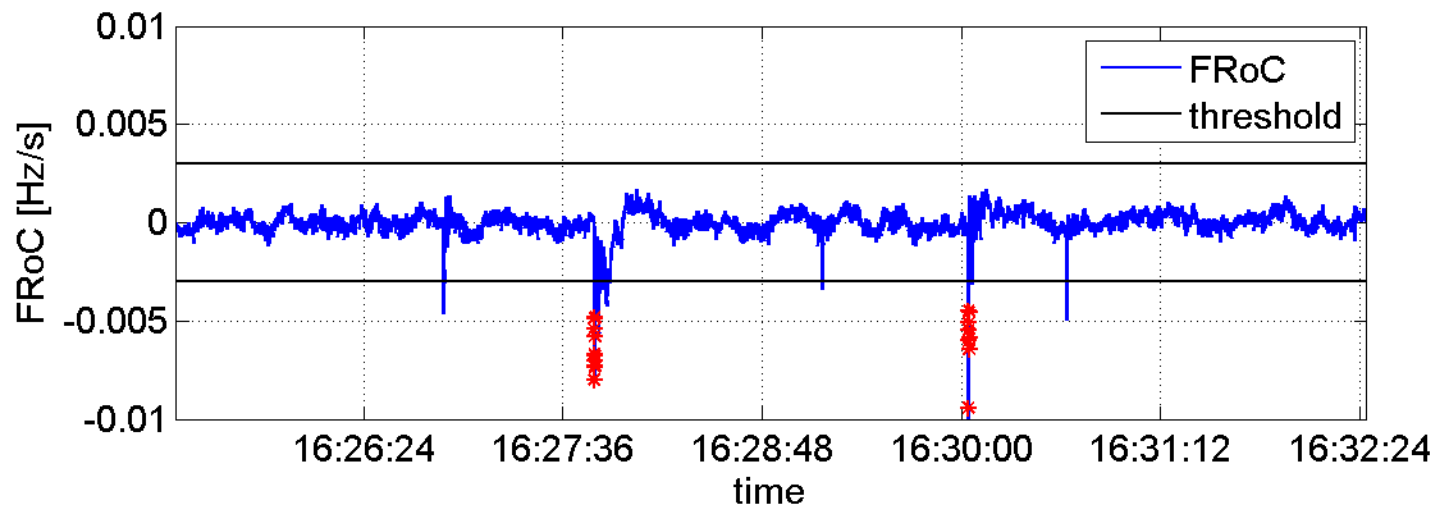
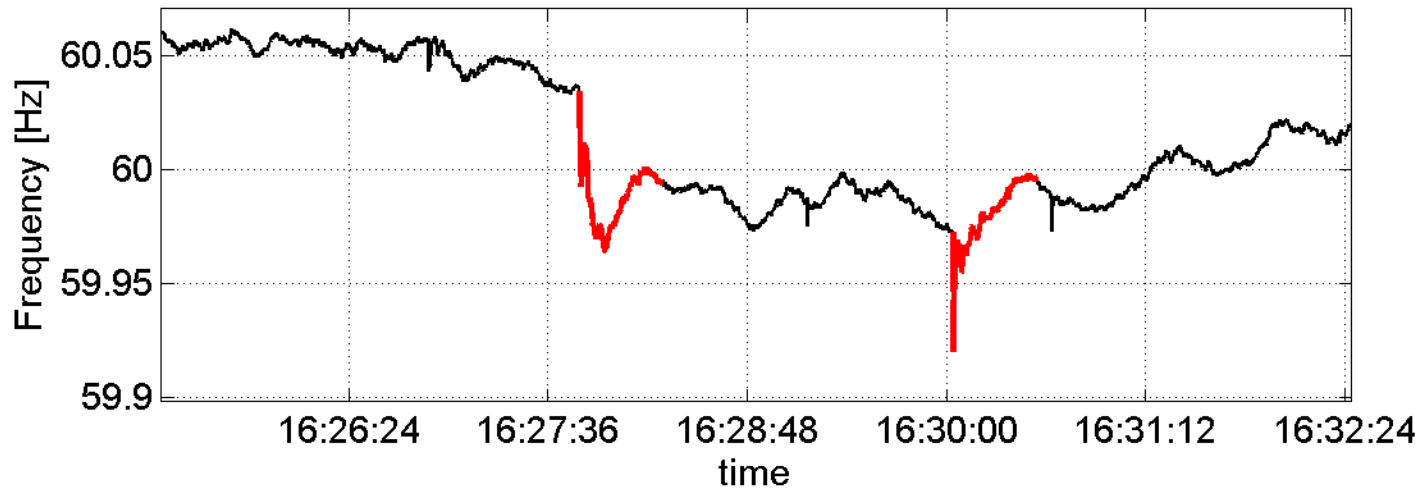
Filtered Rate of Change (FRoC) results - May 30, Jack PMU at UCSD



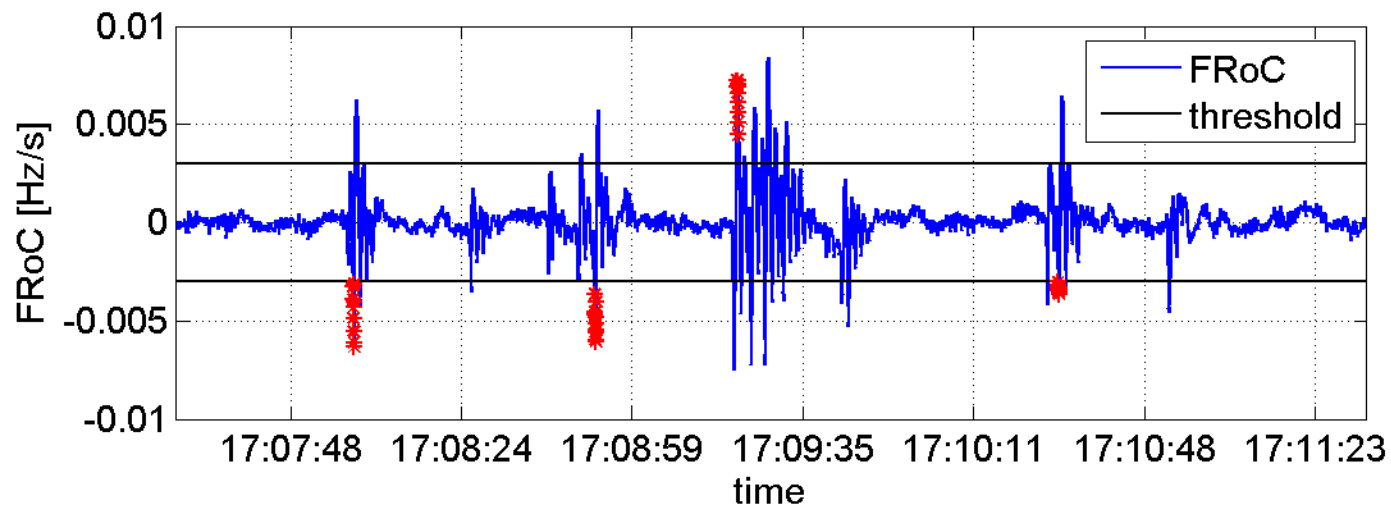
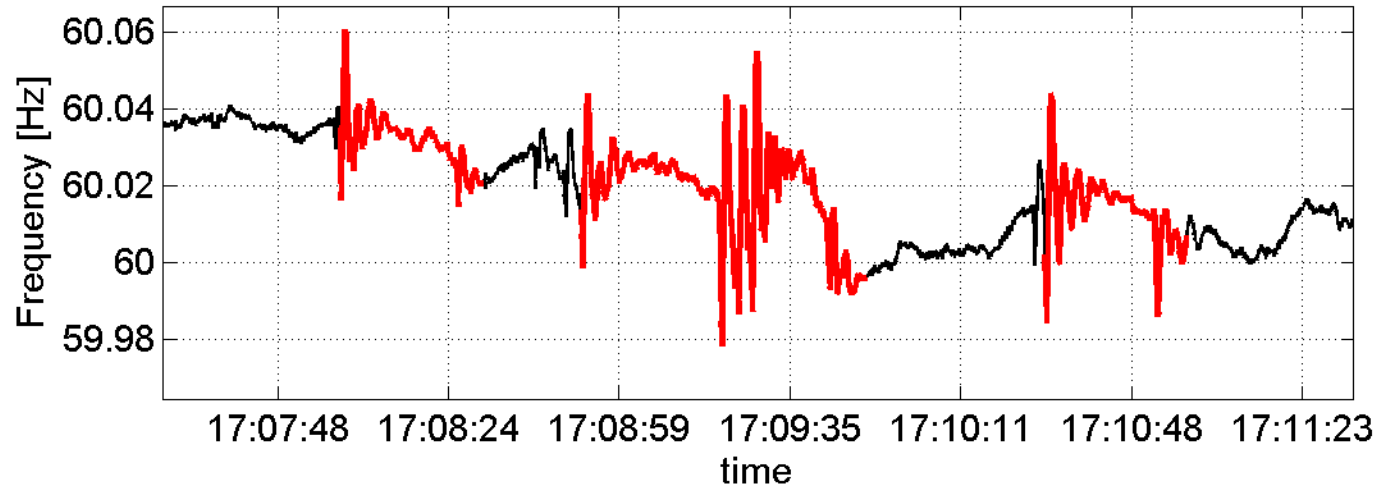
Filtered Rate of Change (FRoC) results - May 30, Jack PMU at UCSD



Filtered Rate of Change (FRoC) results - May 30, Jack PMU at UCSD



Filtered Rate of Change (FRoC) results - May 30, Jack PMU at UCSD



PART2

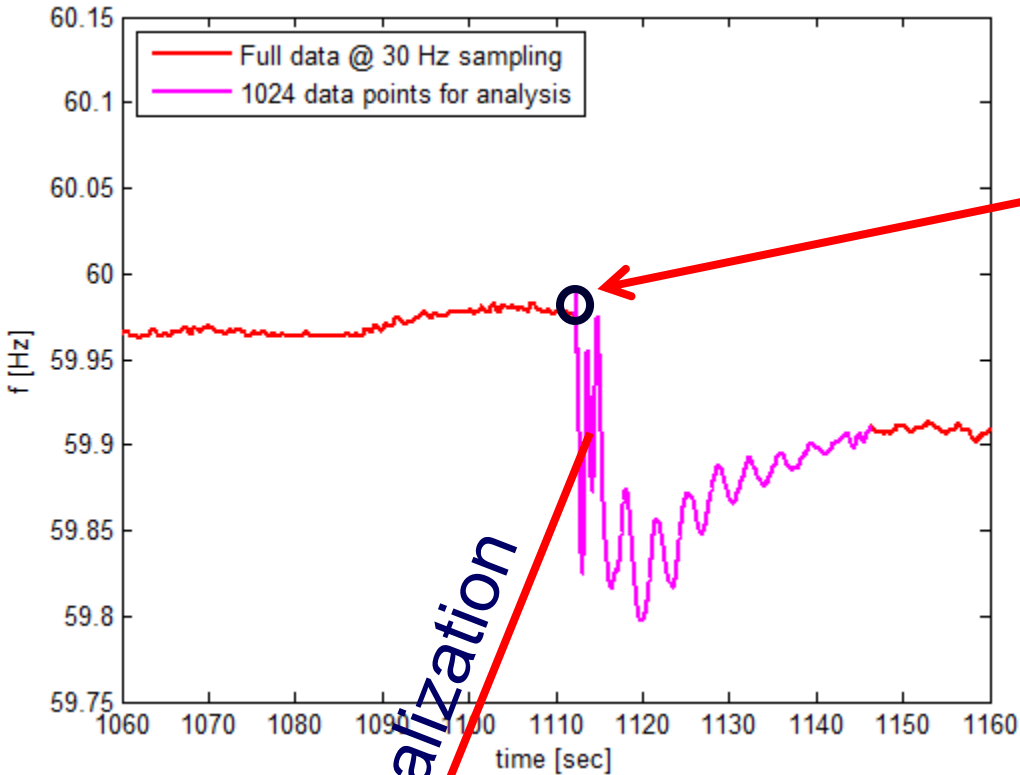
UCSD Microgrid

Ring Down Analysis of Oct. 9 event

Mitigation of events via real-time control

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Disturbance in F3 in JSIS data



detect beginning of event

realization

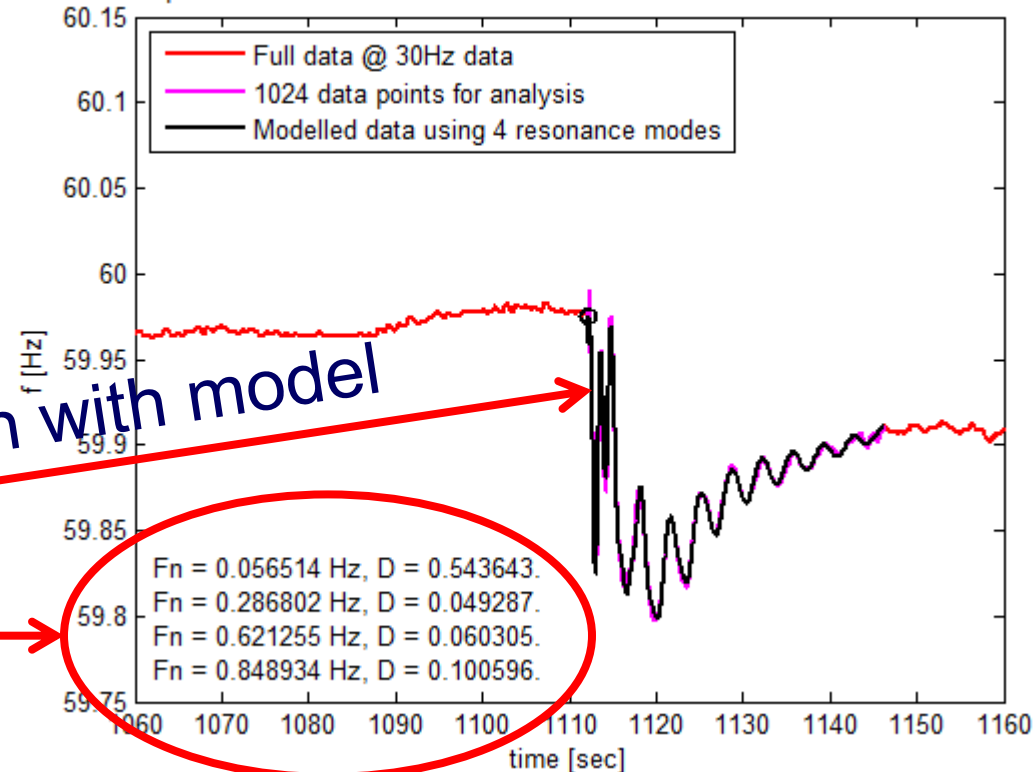
simulation with model

analysis

$$x(t+1) = Ax(t) + Bd(t)$$

$$F(t) = Cx(t)$$

Comparison of actual disturbance in F3 in JSIS data and simulated disturbance



Approach:

- Assume observed event in frequency $F(t)$ is due to a deterministic system

$$x(k+1) = Ax(k) + Bd(k)$$

$$F(k) = Cx(k)$$

Discrete-time model

where (unknown) input $d(t)$ can be 'impulse' or 'step' or 'known shape'

- Store a **finite number of data points** of $F(t)$ in a special data matrix **H**
- Inspect rank of (null projection on) **H**: determines **# modes**
- Compute matrices **A, B and C** via **Realization Algorithm**.
- Extension of Ho-Kalman, Kung algorithm. Miller, de Callafon (2010)
- Applicable to **multiple time-synchronized measurements!** (multiple PMUs)

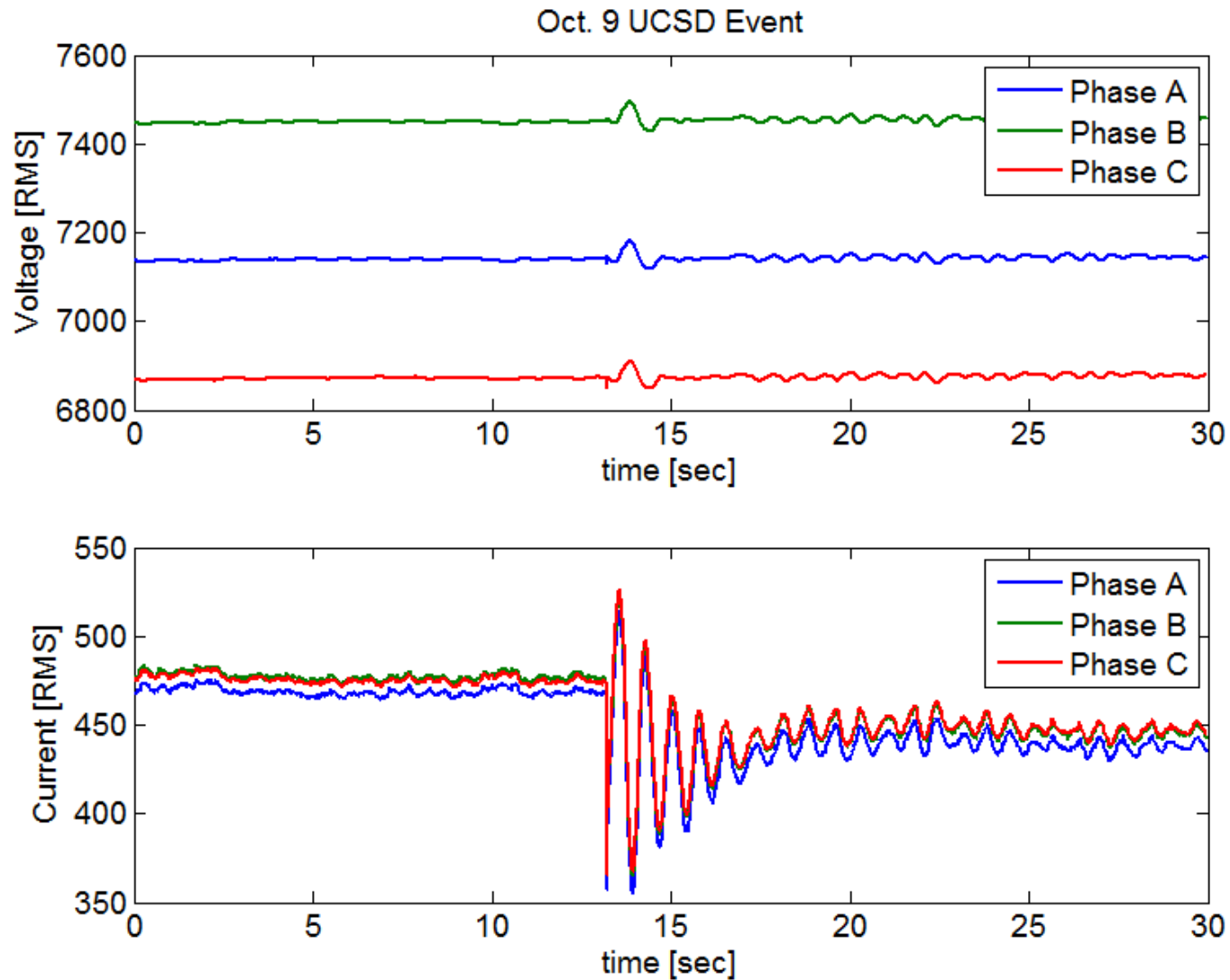
End Result:

- Dynamic model (state space model) can be used for
 - **Simulation**: simulate the disturbance data
 - **Analysis**: Compute **resonance modes and damping** (from eigenvalues of A)

Measurements from SEL breaker at 12kV 3 phase line (6.9kV phase to phase)

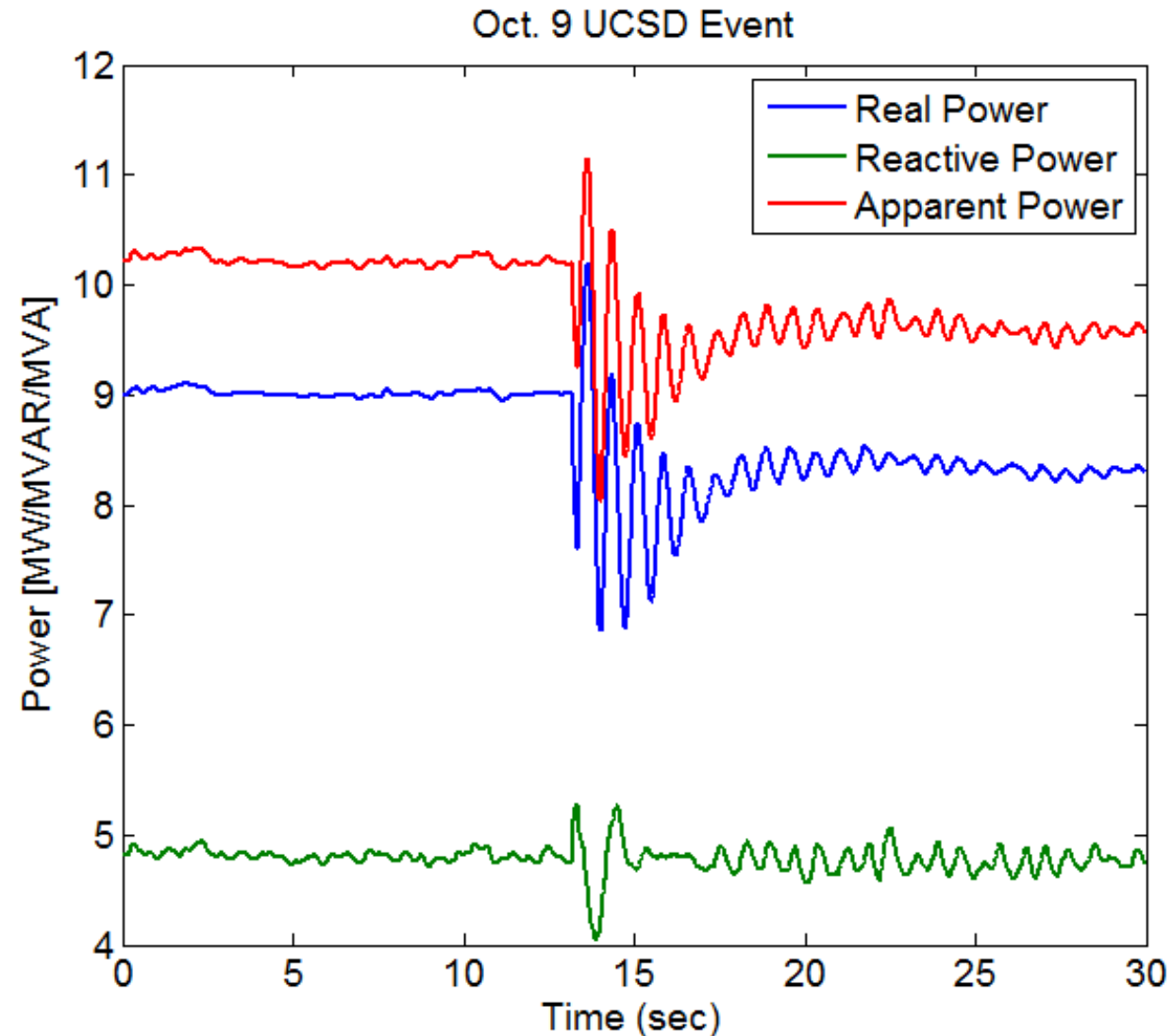
- RMS Voltage and Current of 3 phases
- Real Power
- Apparent Power

Disturbance on 3 phase network



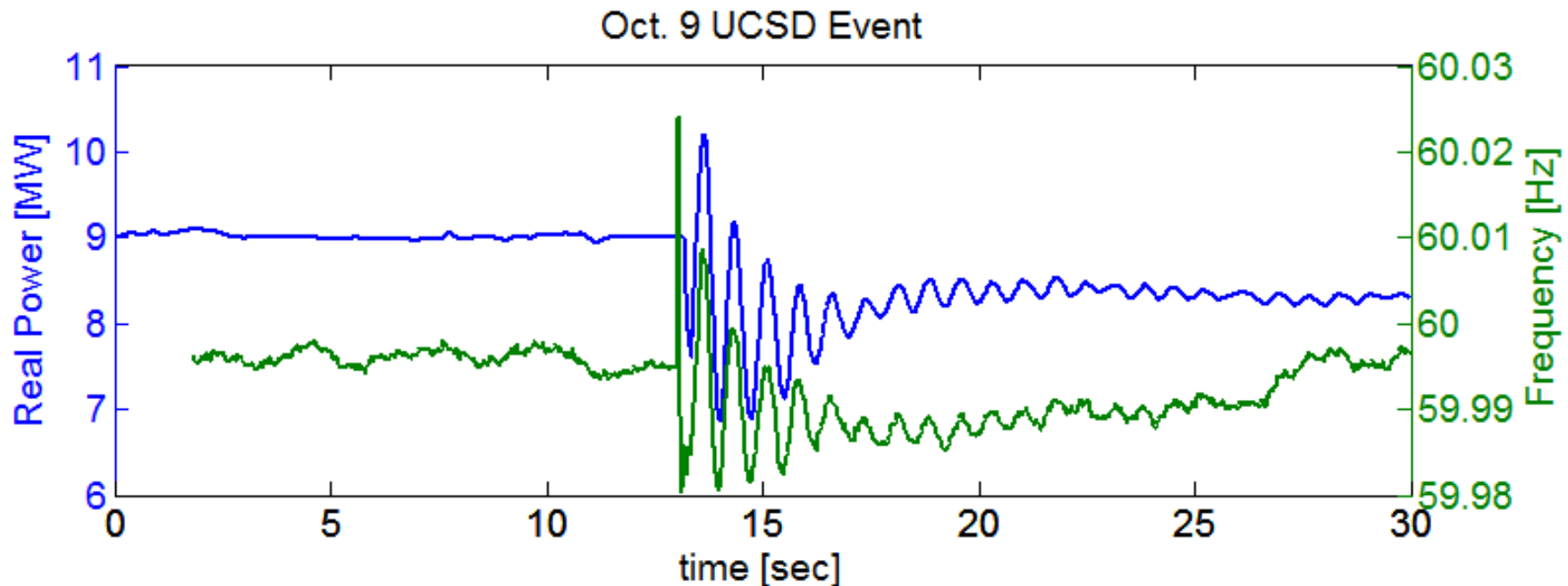
Measurements from
SEL breaker at 12kV
3 phase line (6.9kV
phase to phase)

- RMS Voltage and Current of 3 phases
- Real Power
- Apparent Power



Main conclusions from Measurements from SEL breaker:

- Sustained oscillations in 3 phase V and I mostly due to reactive power.
- Real power oscillations dampen out faster
- (time adjusted) Frequency show similar dynamics as Real Power:



Realization Algorithm:

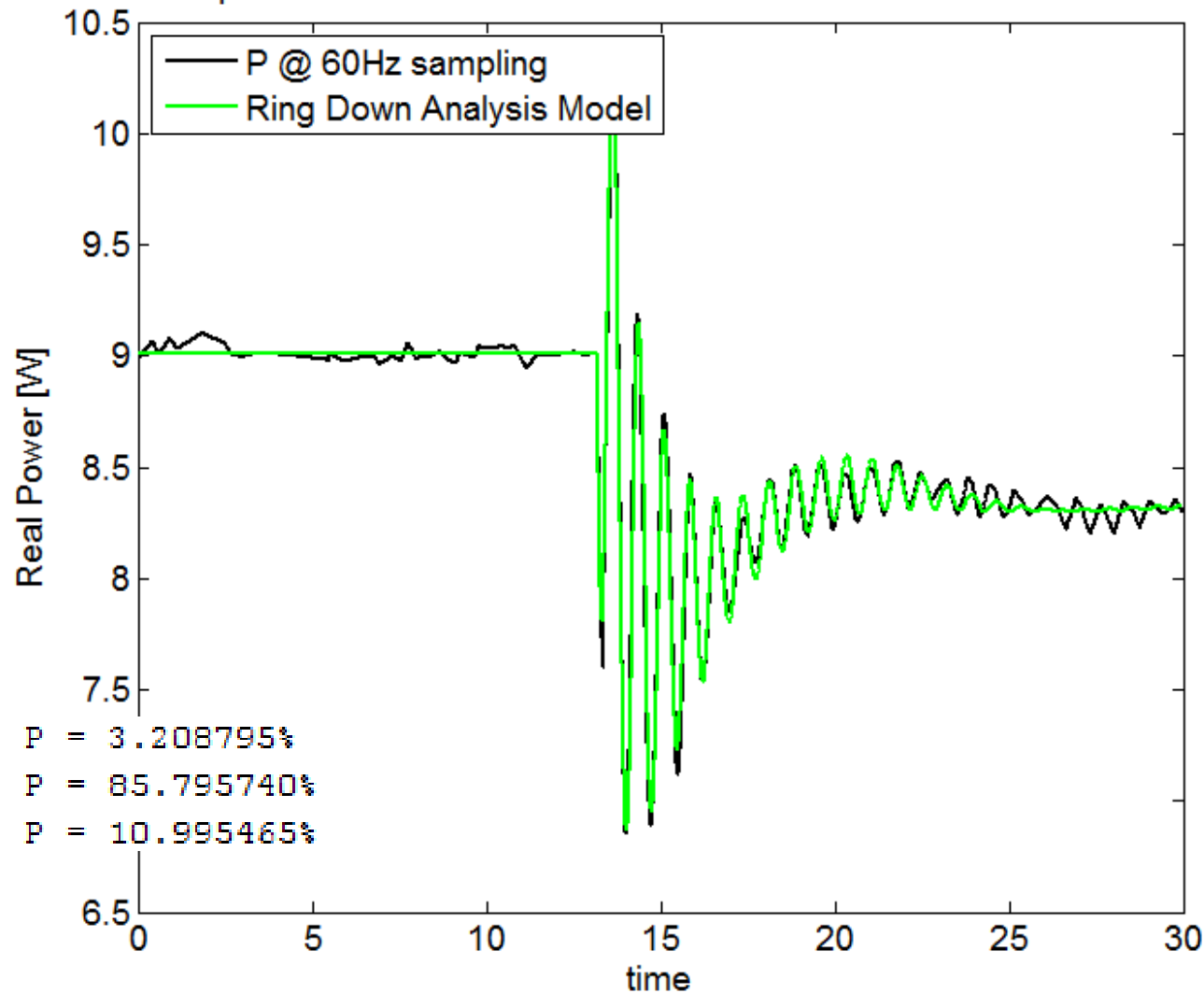
excellent fit of
oscillation/damping

Modeled
frequencies F_n ,
damping D and
model participation P :

$F_n = 0.094653$ Hz, $D = 0.450955$, $P = 3.208795\%$
 $F_n = 1.353568$ Hz, $D = 0.044507$, $P = 85.795740\%$
 $F_n = 1.461354$ Hz, $D = 0.026519$, $P = 10.995465\%$

Mode around 1.4Hz
less than 5% damping, 85% participation

Comparison of actual disturbance in data and simulated disturbance

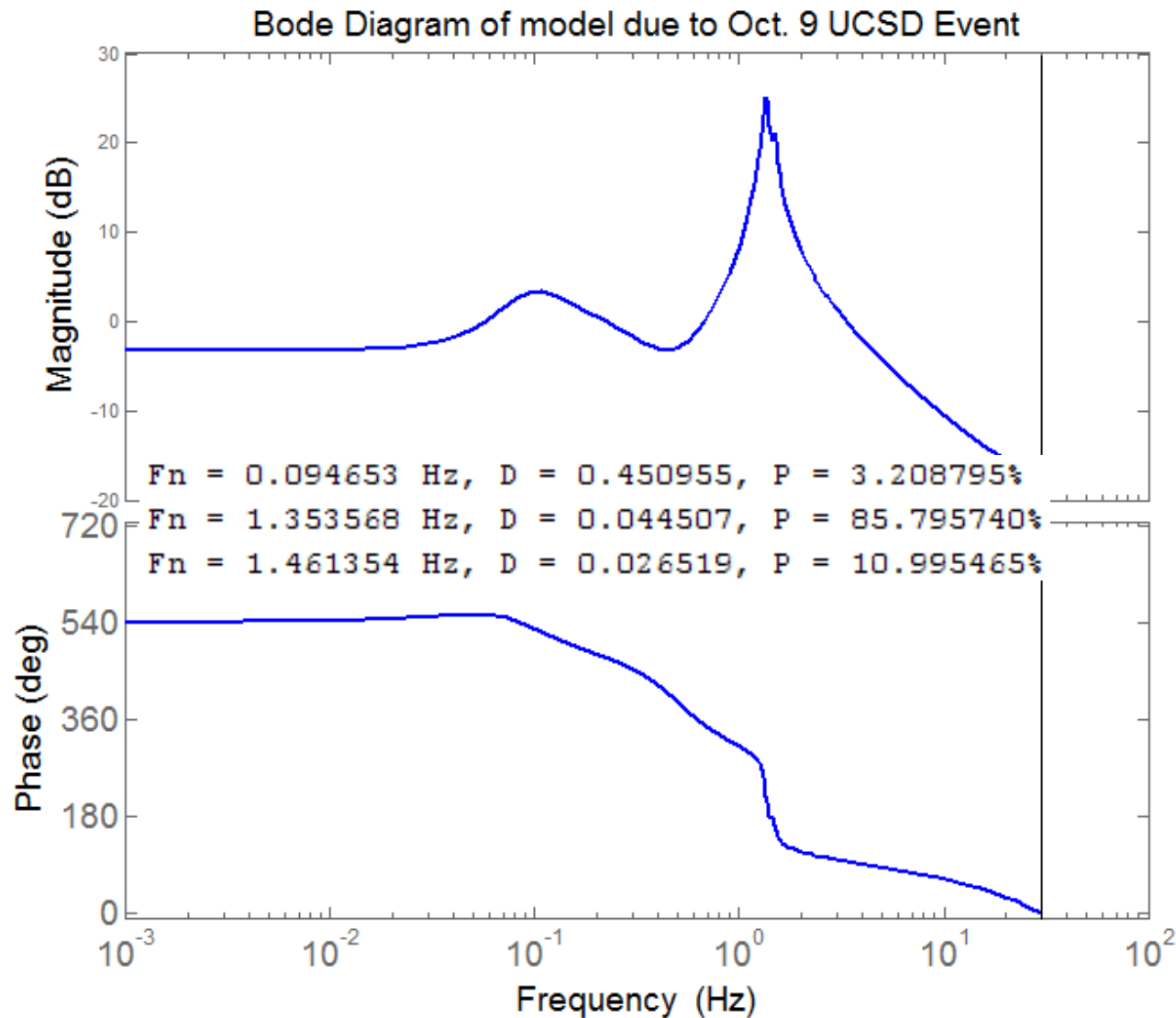


Dynamic model found by realization in Bode plot (frequency domain)

Observe large resonance frequency around 1.4Hz

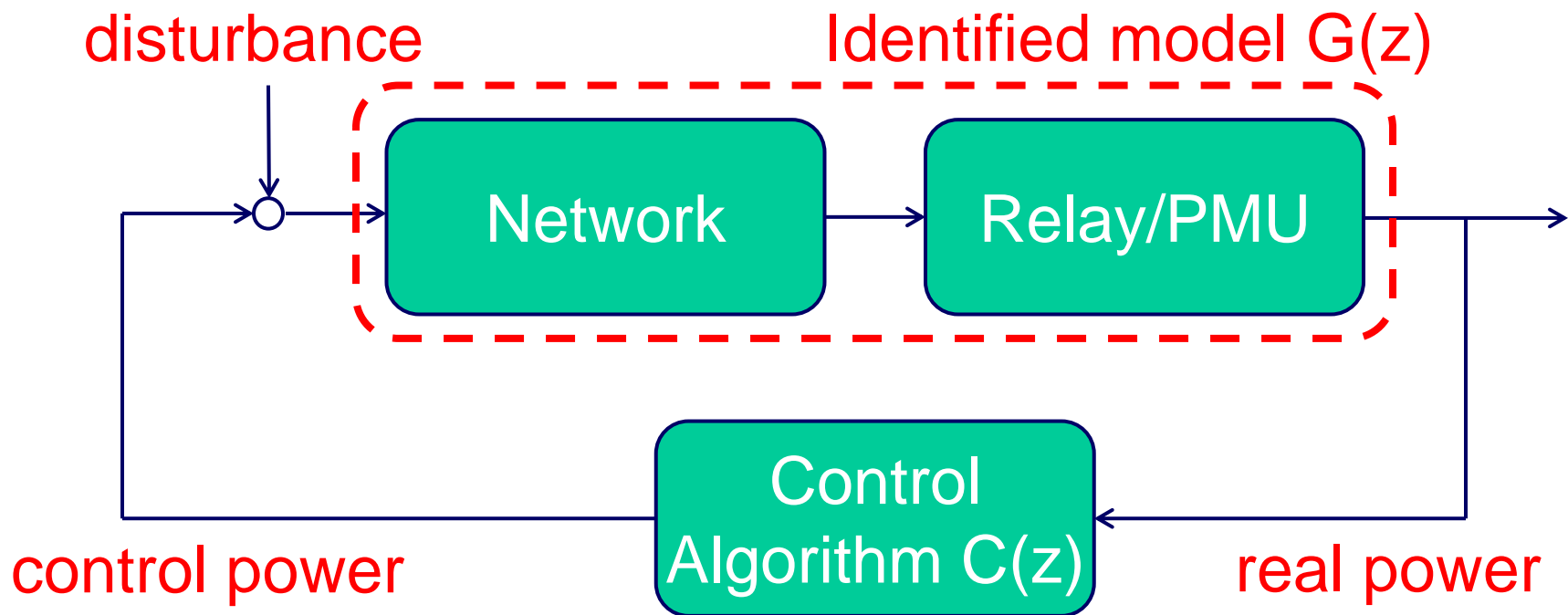
MITIGATION

Control/damping of 1.4Hz oscillation



MITIGATION

Control/damping of 1.4Hz oscillation via Real Power control:



- What is the **control algorithm**?
- How much **control power** is needed to dampen oscillation?

Identified Discrete-Time Model $G(z)$:

$$G(z) = \frac{-0.2791 z^6 + 1.677 z^5 - 4.204 z^4 + 5.63 z^3 - 4.249 z^2 + 1.713 z - 0.2882}{z^7 - 6.89 z^6 + 20.39 z^5 - 33.58 z^4 + 33.26 z^3 - 19.8 z^2 + 6.564 z - 0.9344}$$

Proposed control algorithm $C(z)$ that has the following shape:

$$C(z) = K \frac{z - 1}{(z - a)(z - b)}$$

- Discrete-time differentiator (to add damping + reduce low frequency control)
- Two poles (a,b) to limit bandwidth
- Gain K to adjust power gain

Choice of control parameters
K, and b in

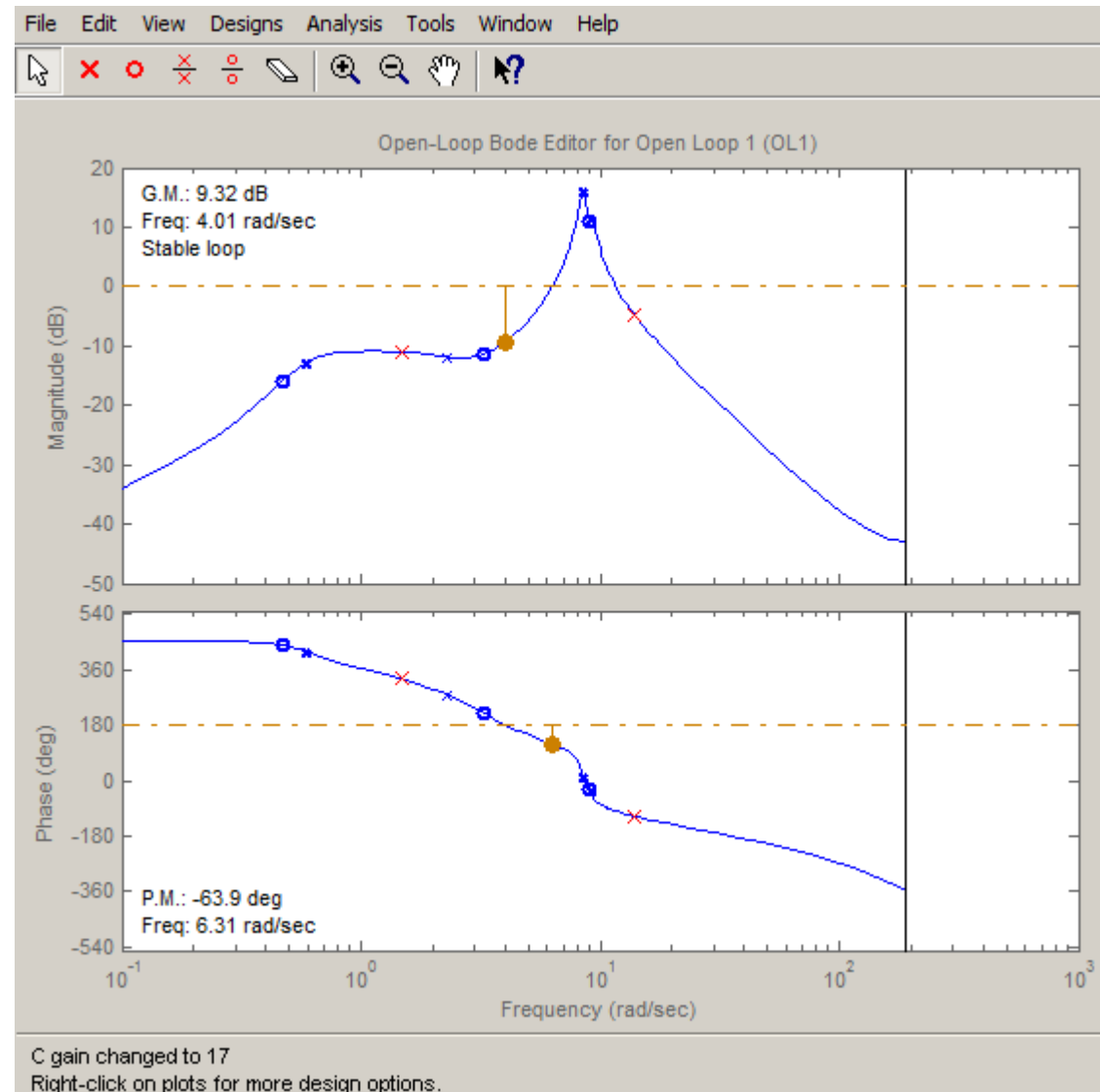
$$C(z) = K \frac{z - 1}{(z - a)(z - b)}$$

via **loop shaping** tool

Shape Bode plot of
 $L(z) = G(z)C(z)$

See direct effect of:

- Damping
- Stability
- Control signal

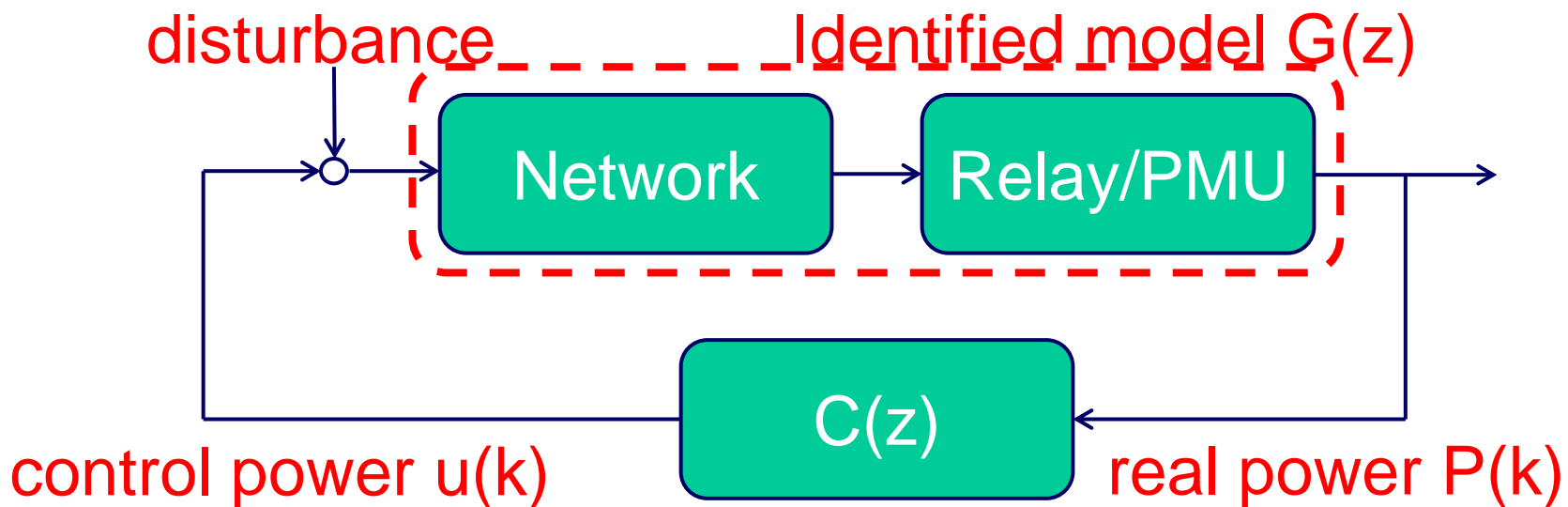


End result of control design:

$$C(z) = K \frac{z - 1}{(z - a)(z - b)}, K = 0.085211, a = 0.9757, b = 0.7933$$

Resulting discrete control algorithm:

$$u(k) = 0.0852 \cdot P(k - 1) - 0.0852 \cdot P(k - 2) + 1.7690 \cdot u(k - 1) - 0.7740 \cdot u(k - 2)$$



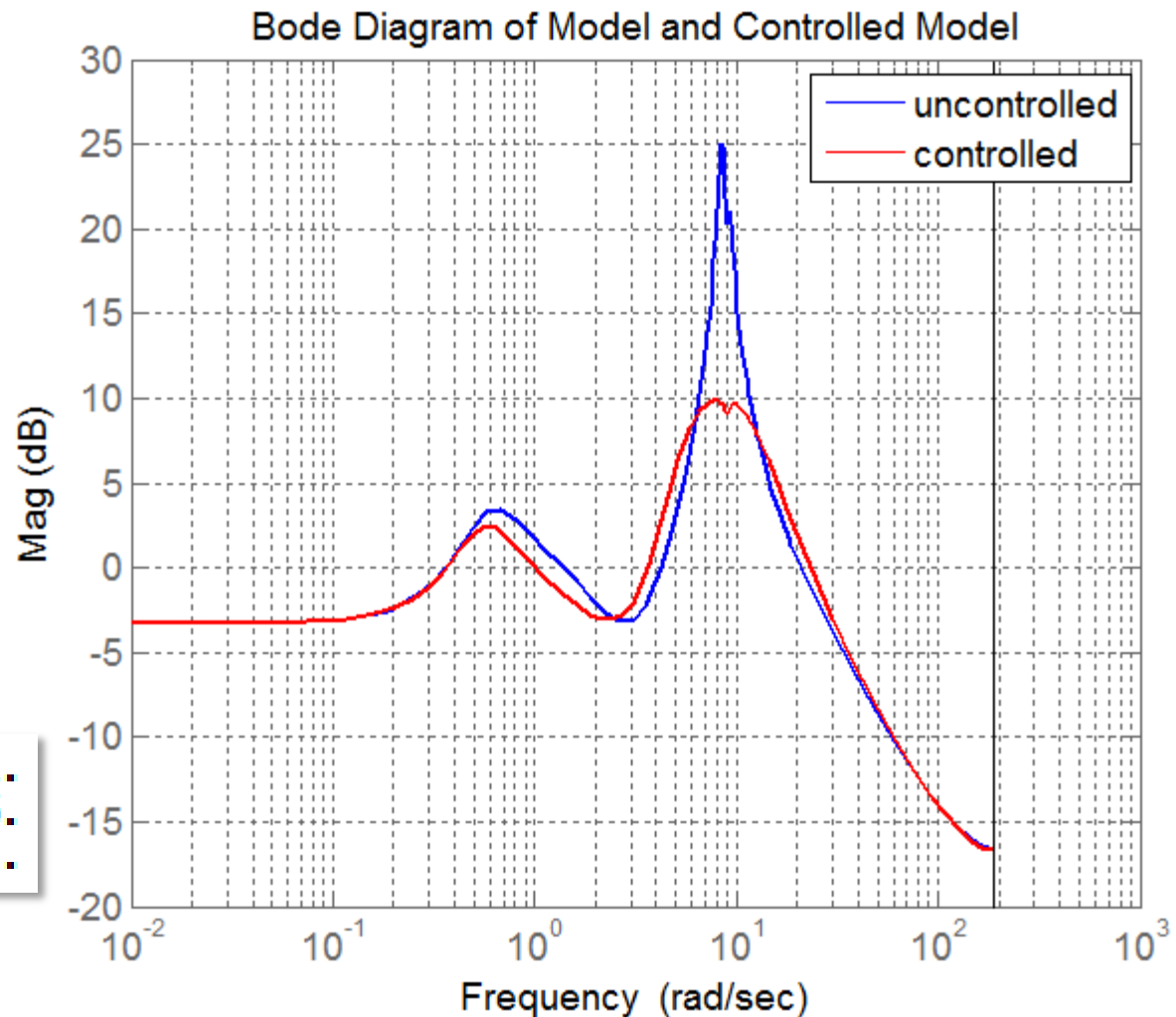
Effect of Control Algorithm:

Damping of UCSD microgrid:

```
Fn = 0.094653 Hz, D = 0.450955.
Fn = 1.353568 Hz, D = 0.044507.
Fn = 1.461354 Hz, D = 0.026519.
```

Damping of controlled UCSD microgrid:

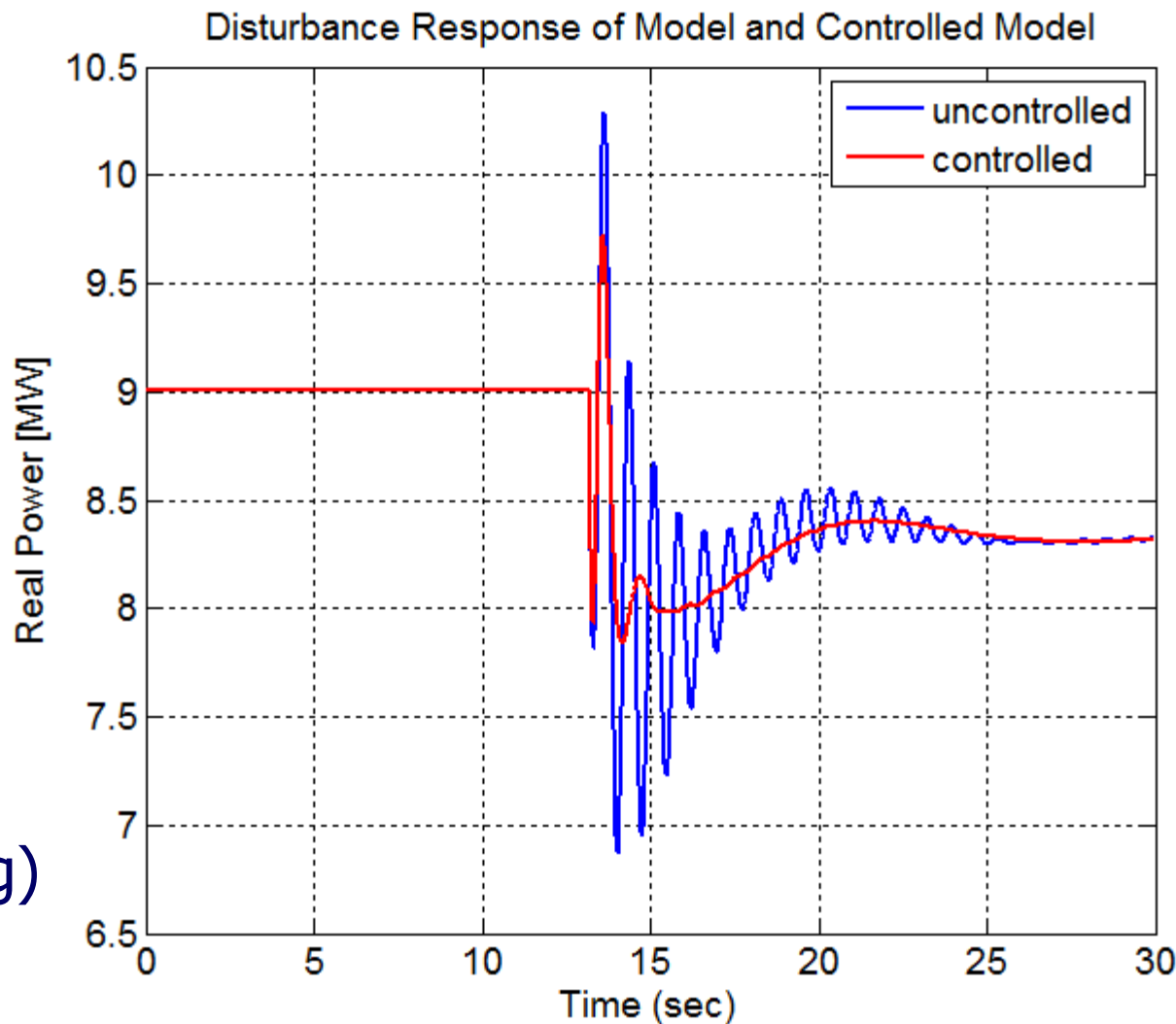
```
Fn = 0.089560 Hz, D = 0.445131.
Fn = 0.904540 Hz, D = 0.415226.
Fn = 1.771599 Hz, D = 0.502977.
```



Slight change in resonance modes, ten-fold increase in damping!

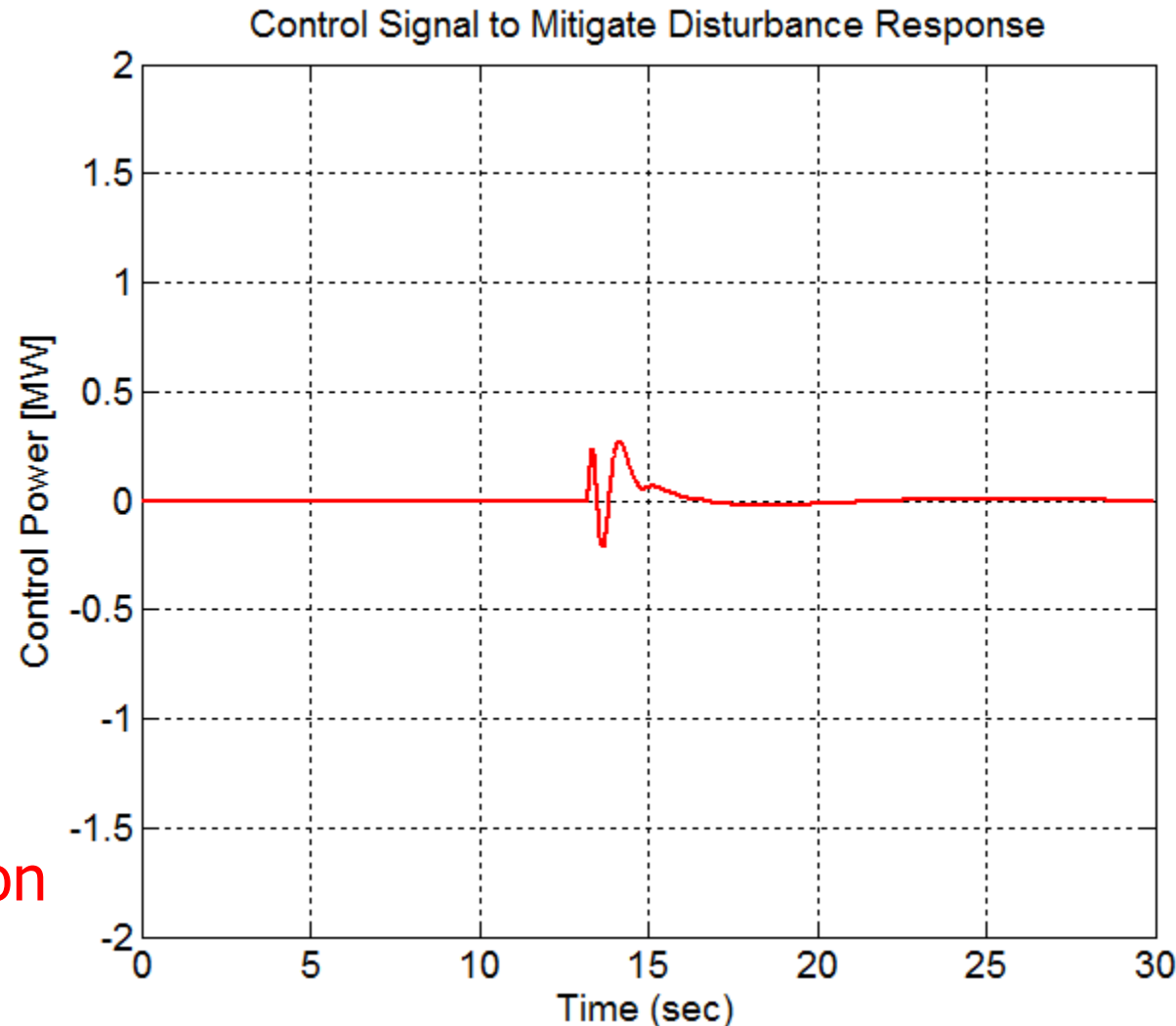
Effect of Control Algorithm:

- Disturbance effect still present (unavoidable)
- Control algorithm does mitigate disturbance faster!
- Less oscillations in microgrid (better damping)
- How much control power needed?



Effect of Control Algorithm:

- For comparison, control power plotted at same scale as a disturbance in real power
- Disturbance almost +/- 2MW
- Control power only +/- 0.25MW for mitigation
- Results scale with size of disturbance and increase of damping



Reducing control effort to +/- 125KW still works, but:

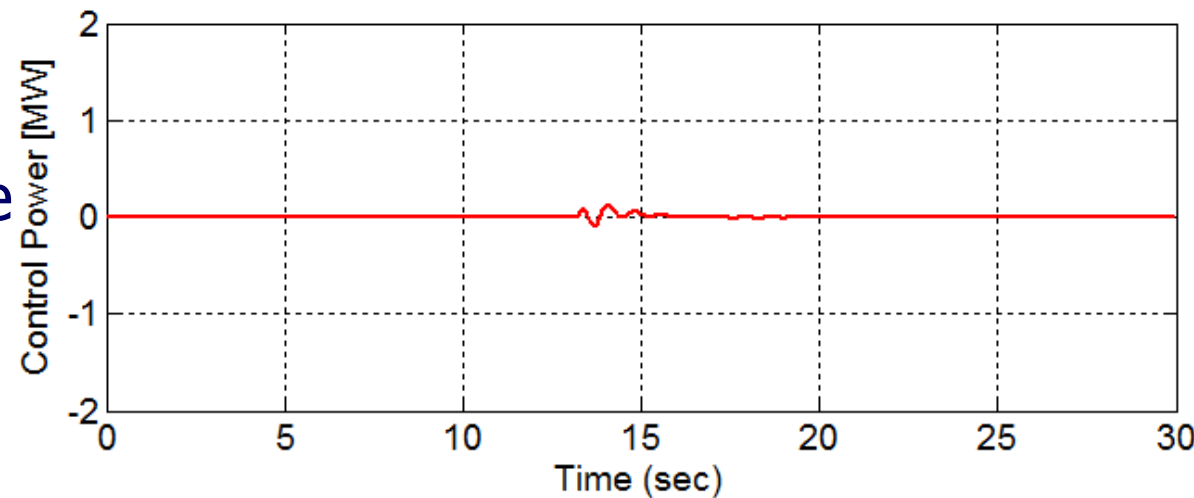
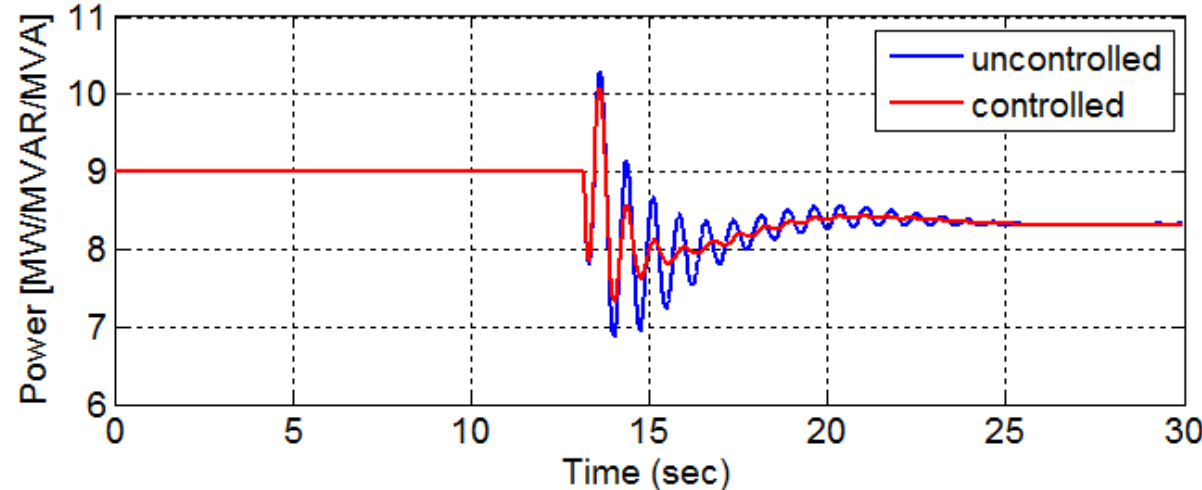
- Damping cannot be influenced that much

```
Fn = 0.092977 Hz, D = 0.448233.
Fn = 1.349573 Hz, D = 0.132450.
```

- Still acceptable to improve dynamics of microgrid

- Control power only +/- 125KW for mitigation

Control Signal (bottom) to Mitigate Disturbance (top)



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- **Automatically detect** when a disturbance/transient event occurs
- **Automatically estimate** Frequency, Damping and Dynamic Model.

Main Features:

- **Automatically detect event:**
 - **Predict** ambient Frequency signal "one-sample" ahead
 - Observe when prediction deviates for **event detection via FRoC signal**
- **Automatically estimate:**
 - **# of modes** of oscillations in measured disturbance
 - Estimate **frequency and damping** of the modes
 - Put results in **dynamic mode**
- All done in real-time!
- Note: resulting dynamic model can be used for feedback control design to mitigate event!

■ Thank you