

Early warning signs of instability in the statistical properties of PMU data



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NY city, Nov. 9, 1965
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US Northeast and Canada
August 14, 2003
50 million people



A wide-angle, nighttime photograph of a city skyline, likely San Francisco, viewed from across a body of water. The buildings are illuminated with warm yellow and orange lights, contrasting with the dark blue night sky. The water in the foreground is dark and still.

California, Arizona, Mexico
September 8, 2011
5 million people

Northern India

July 30, 2012: 350 million people

July 31, 2012: 700 million people



Photo: Bikas Das/AP Photo
IEEE Spectrum, Oct. 2012

Bangladesh. 1 November 2014

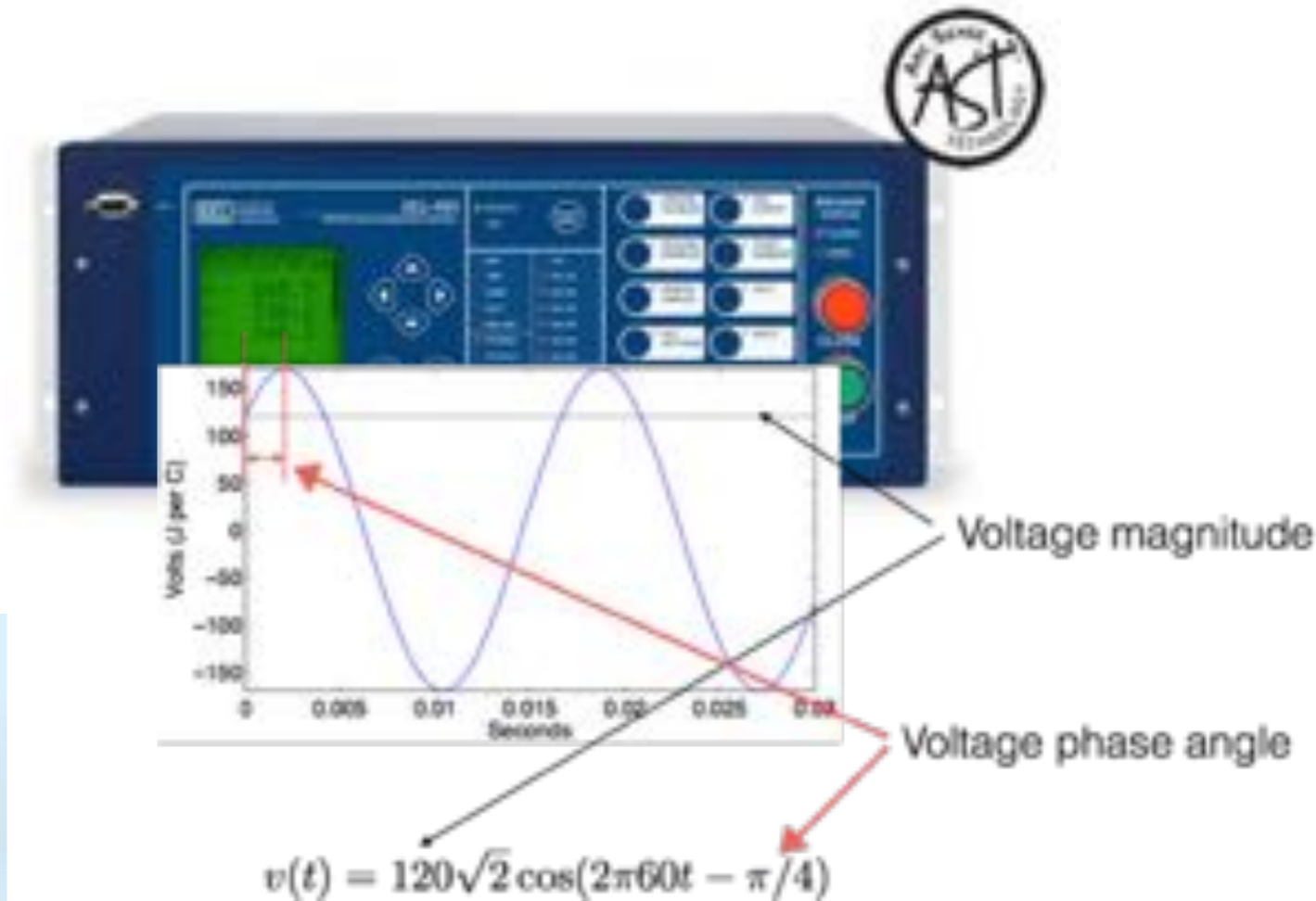
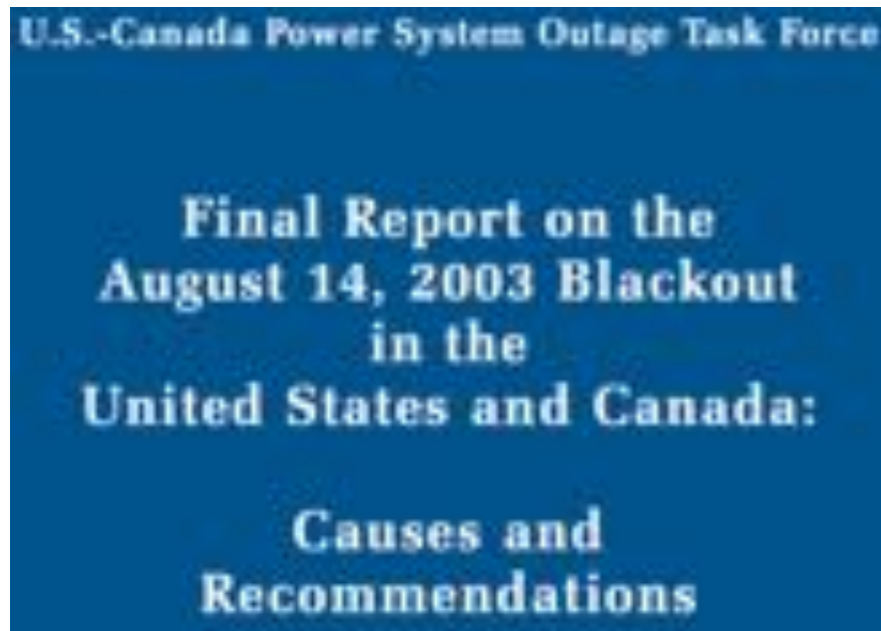


Officials said it would take at least 12 hours to repair the system and restore power to the capital Dhaka [AP]

Washington DC, April 7, 2015



Situational Awareness



Inadequate Situational Awareness

The 2003 Blackout Report stated, “A principal cause of the August 14 blackout was a lack of situational awareness, which was in turn the result of inadequate reliability tools and backup capabilities.”¹⁰⁹ Similarly, the instant inquiry determined that inadequate real-time situational awareness contributed to the cascading outages. In



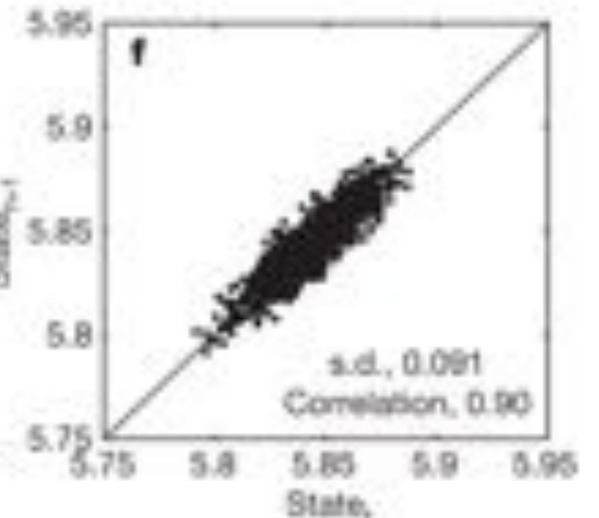
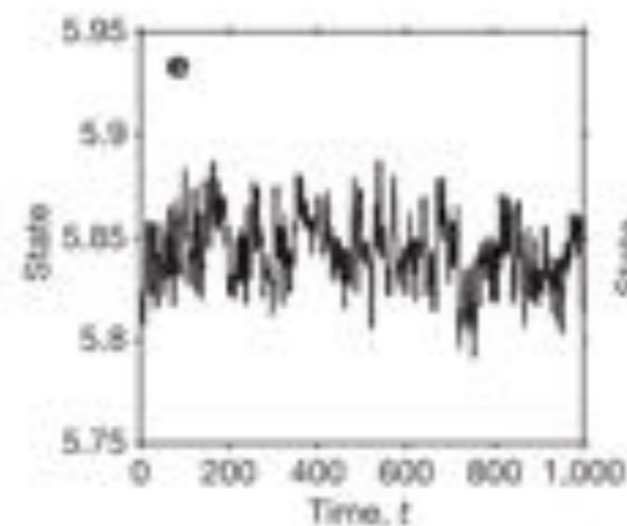
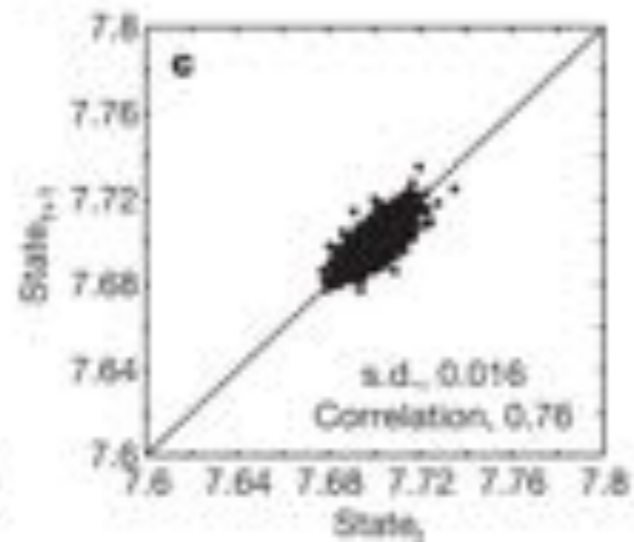
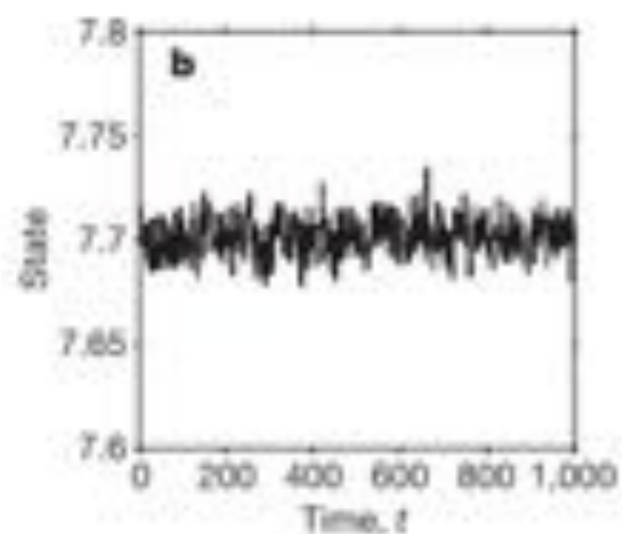
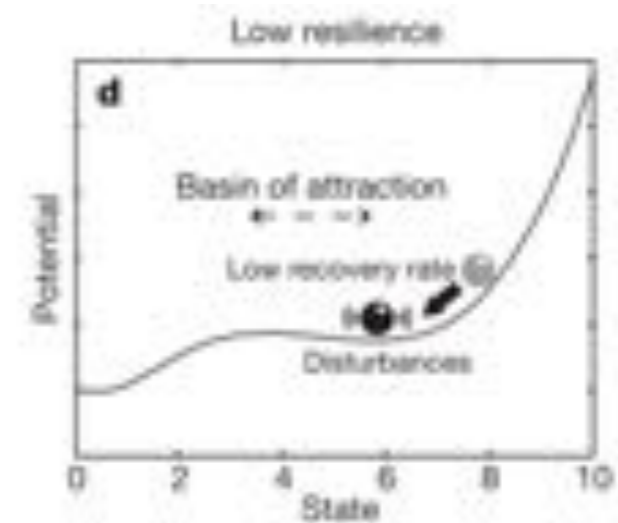
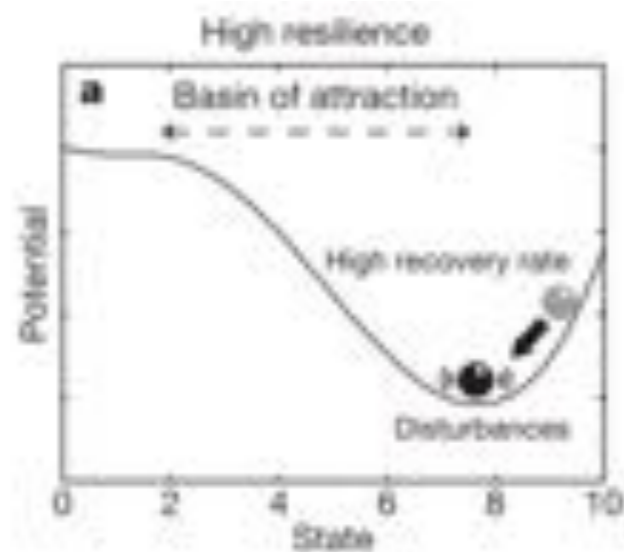
Critical Slowing Down

Vol 461|3 September 2009|doi:10.1038/nature08227

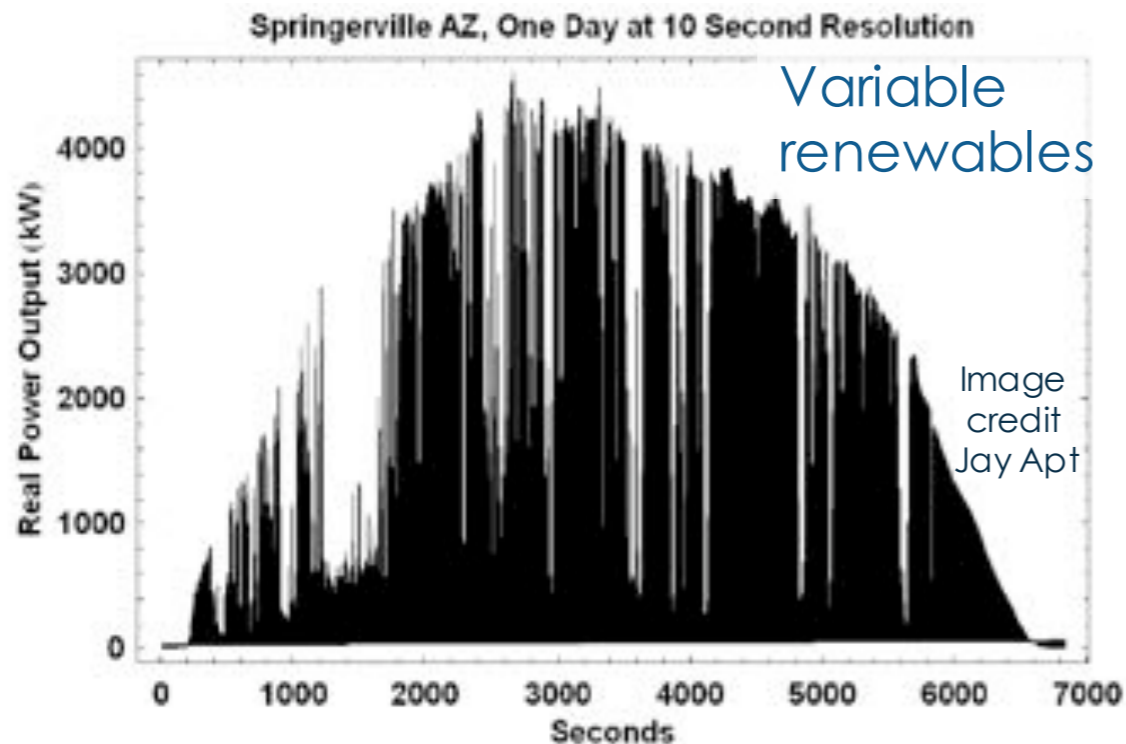
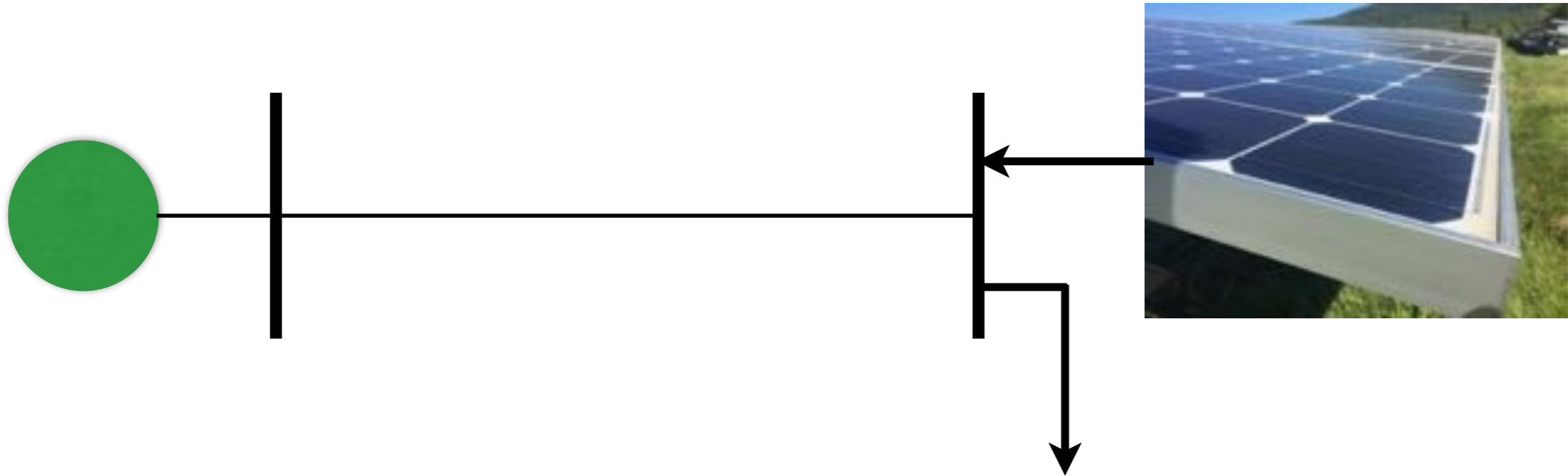
nature

Early-warning signals for critical transitions

Marten Scheffer¹, Jordi Bascompte², William A. Brock³, Victor Brovkin⁵, Stephen R. Carpenter⁴, Vasilis Dakos¹, Hermann Held⁶, Egbert H. van Nes¹, Max Rietkerk⁷ & George Sugihara⁸

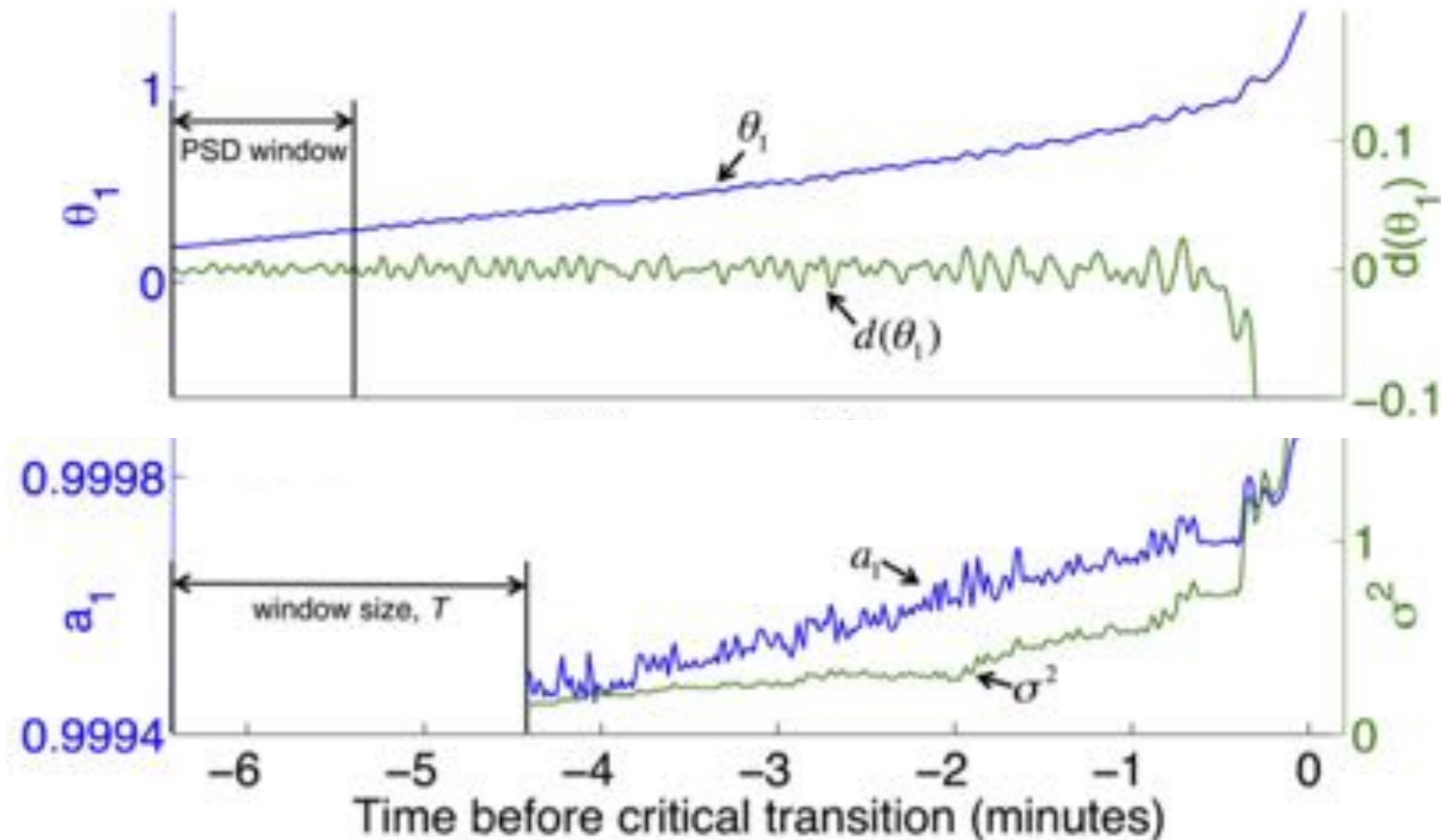


Power systems are constantly “bumped” by randomness



Can that statistical responses to these bumps, measured in PMU data, tell us about the health of the grid?

Statistics can be useful indicators



Cotilla-Sanchez, Hines, Danforth, IEEE Trans Smart Grid, 2012.

See also:

DeMarco and Berge, IEEE Trans on Ckt & Sys, 1987.

Dhople, Chen, DeVille, Domínguez-García, IEEE Trans on Ckt Sys, 2013

Podolsky and Turitsyn, arXiv:1307.4318, Jul. 2013.

Susuki and Mezic, IEEE Trans. Power Syst., 2012 (and others)

How can we find the useful* statistical early warning signs?

**Useful*: A sign that shows up early enough that we might actually be able to do something about it, even if there is measurement noise

Model a power grid using stochastic differential equations

$$\dot{\underline{x}} = f(\underline{x}, \underline{y})$$

Differential equations.
(swing eqs., governors,
exciters, etc.)

$$0 = g(\underline{x}, \underline{y}, \underline{u})$$

Algebraic equations

r.v. for stochastic load perturbations

$$\dot{\underline{u}} = -E\underline{u} + C\xi$$

Loads modeled as Ornstein–Uhlenbeck process

Ind. Gaussian r.v.s, 1% std. dev.

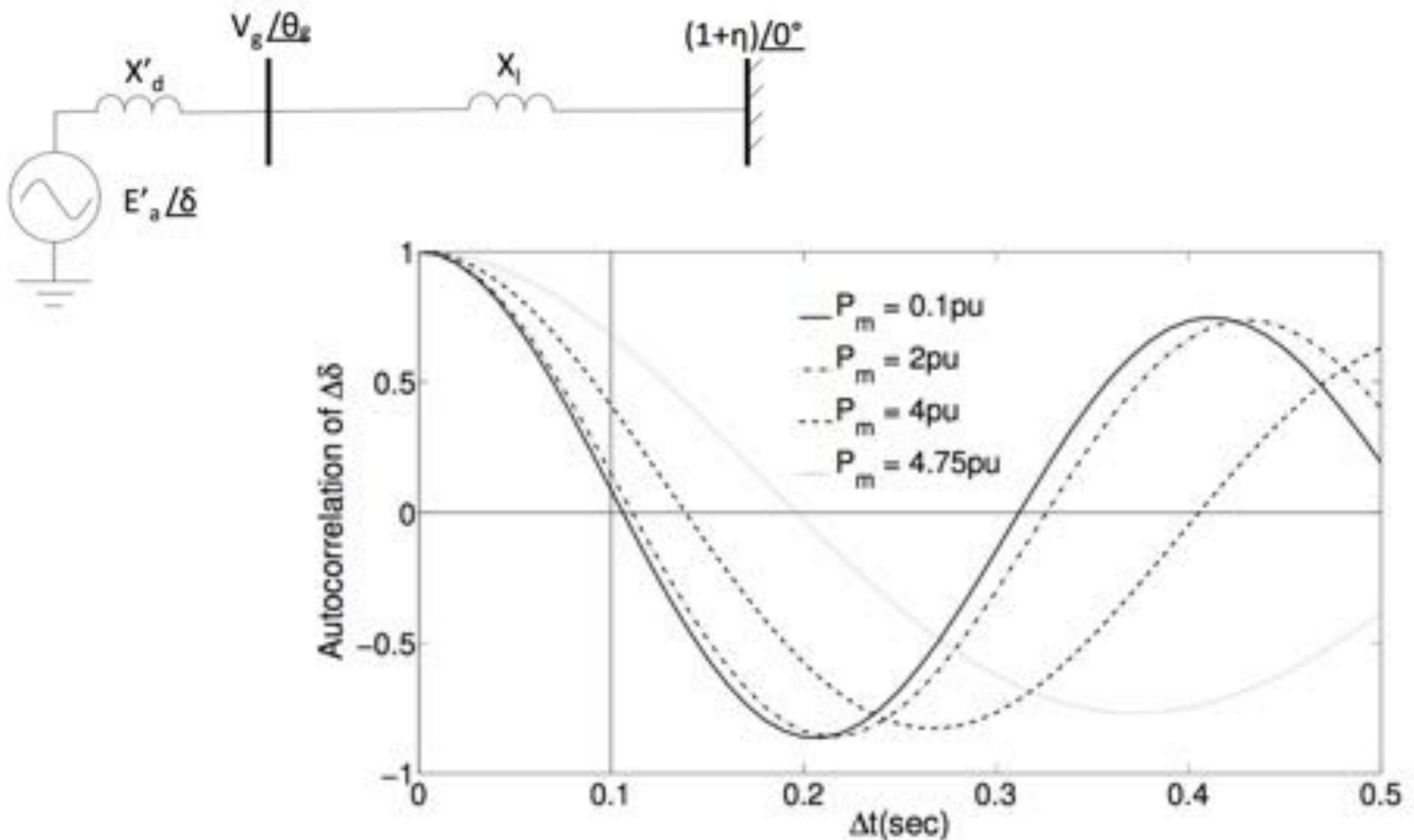
Encodes corr. time of load fluctuations

And solve to find the variance and autocorrelation of voltages and currents

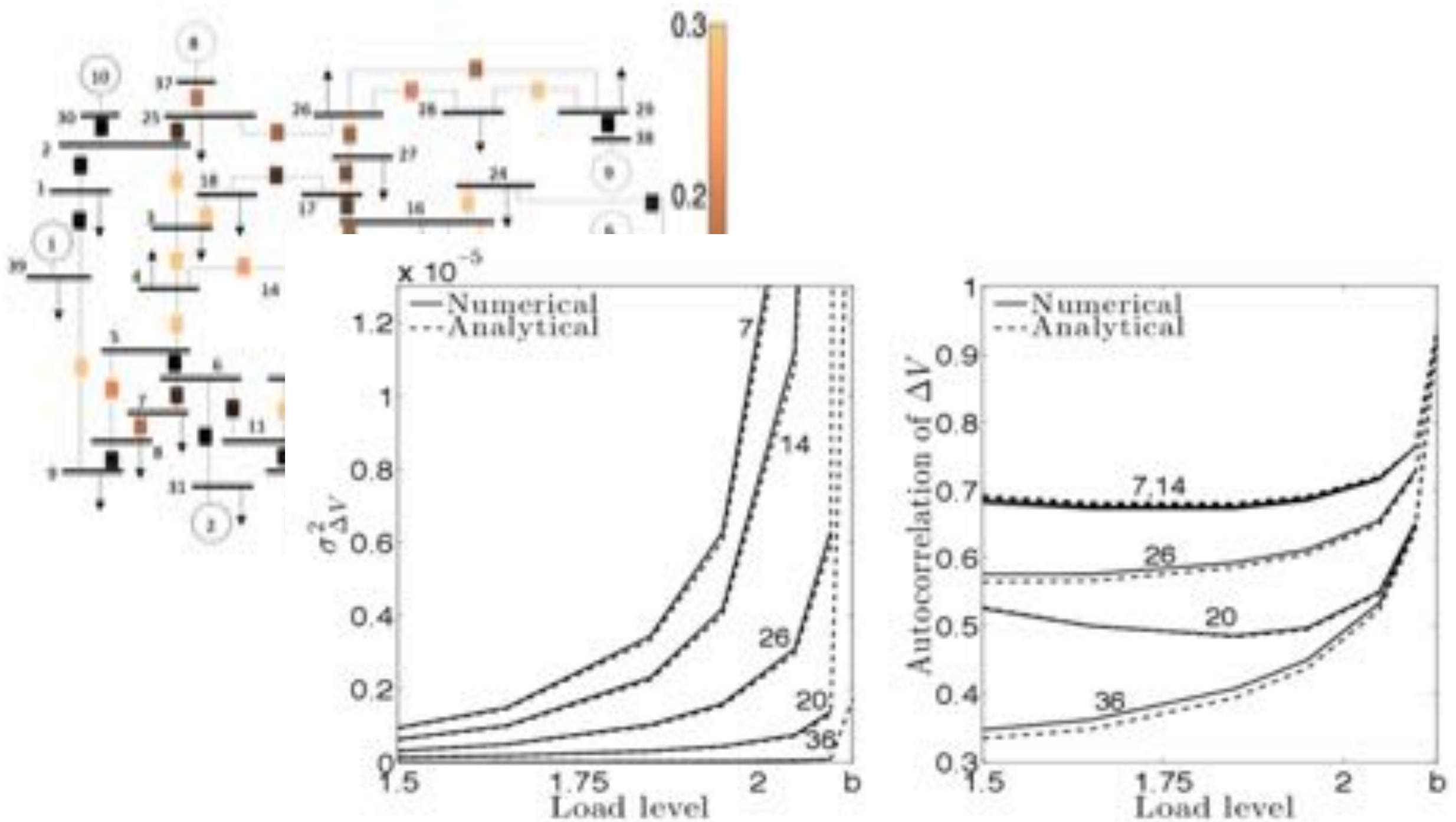


$$A\sigma_{\underline{z}} + \sigma_{\underline{z}}A^T = -BB^T \quad \text{Lyapunov eq.}$$
$$\mathbb{E} [\underline{z}(t) \underline{z}^T(s)] = \exp[-A|t-s|] \sigma_{\underline{z}}$$

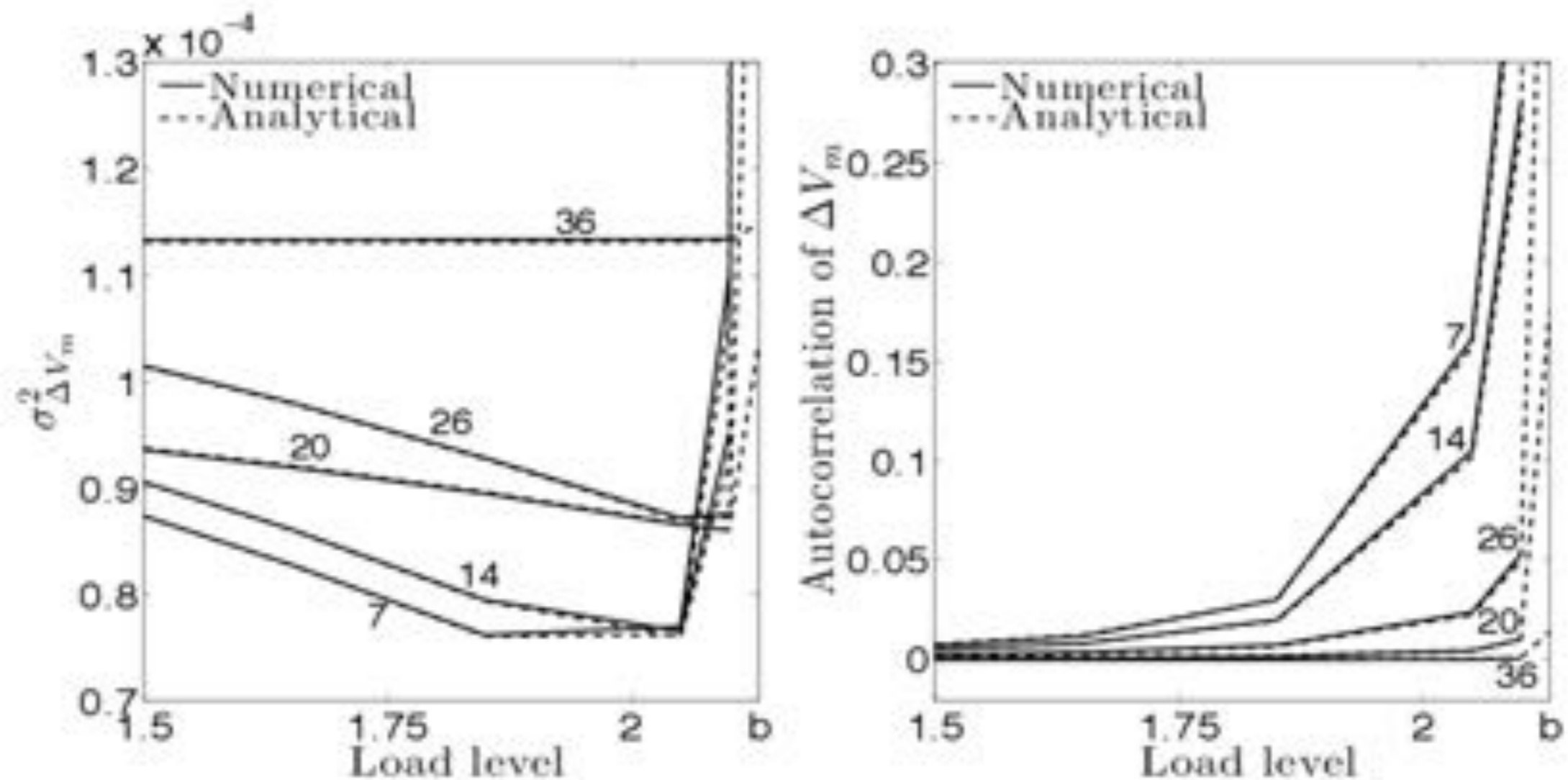
and choose a time delay for autocorrelation measurements



Check to make sure that the analytical and numerical line up



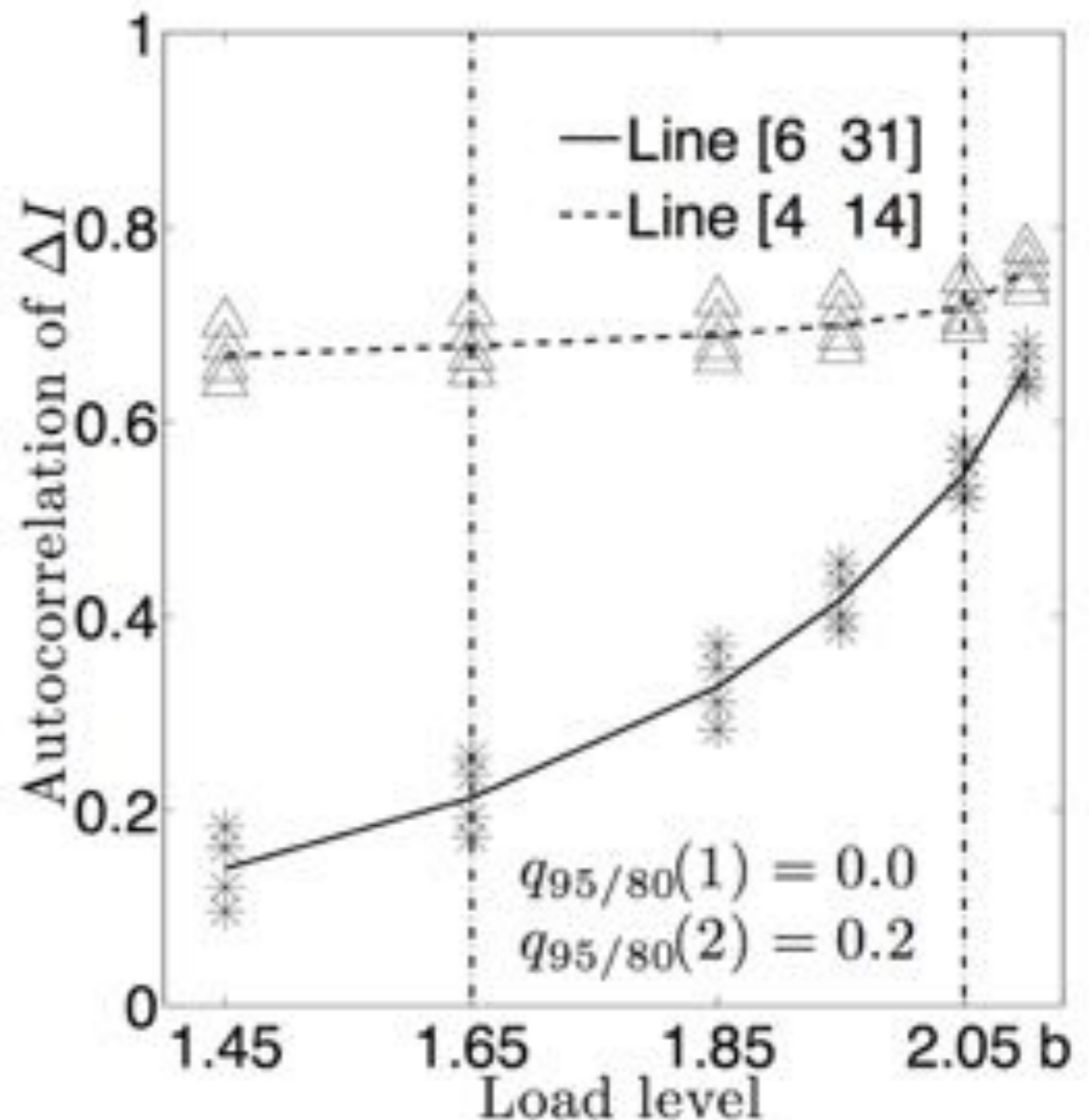
And add measurement noise



Which we can subsequently filter to largely regain our original signal, with the interesting side-effect that some of the variance now appears as autocorrelation.

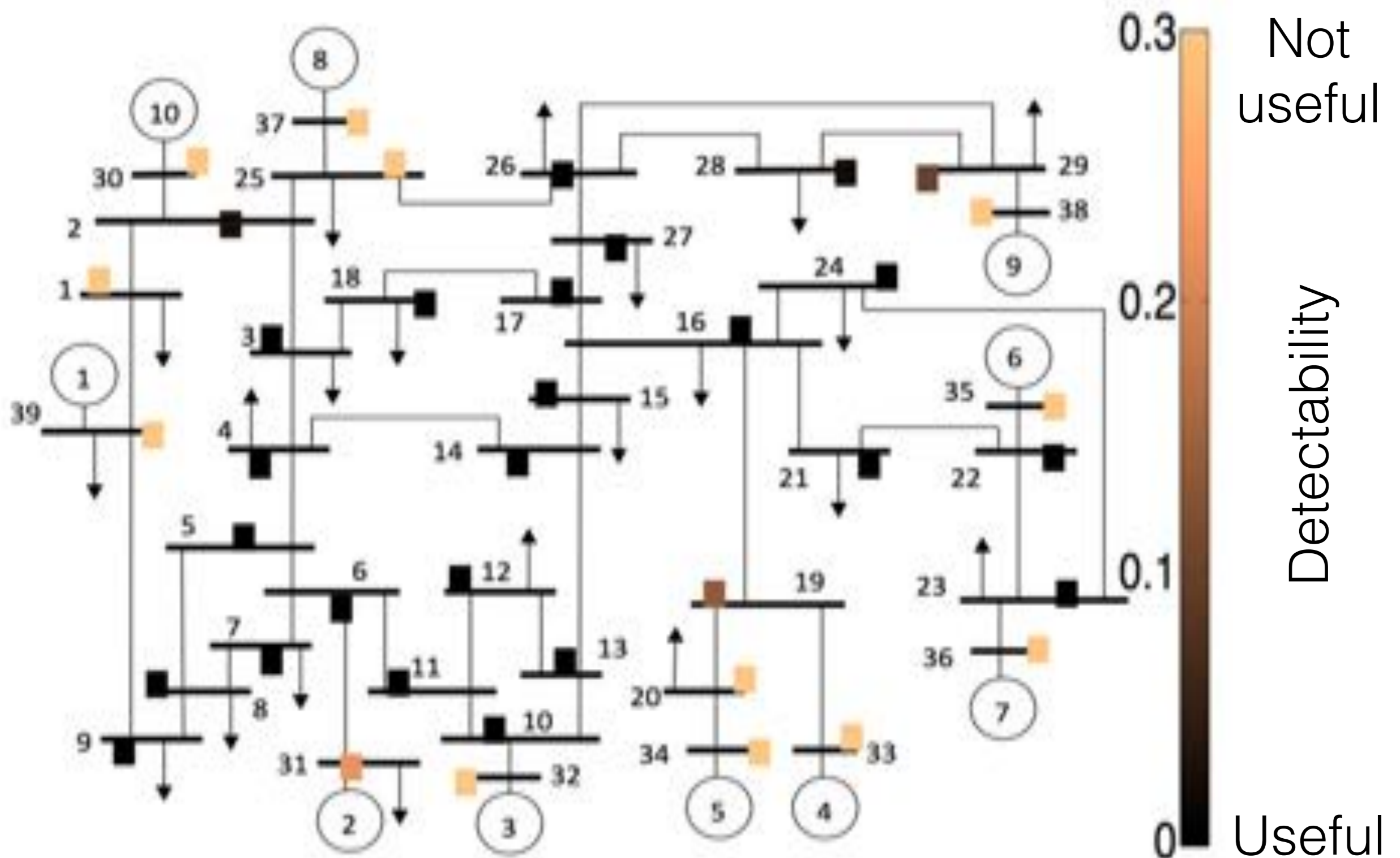
At key locations, we can see clear signs of instability in Autocorrelation and Variance

How do we measure “detectability” to distinguish useful statistical signals from non-useful ones?

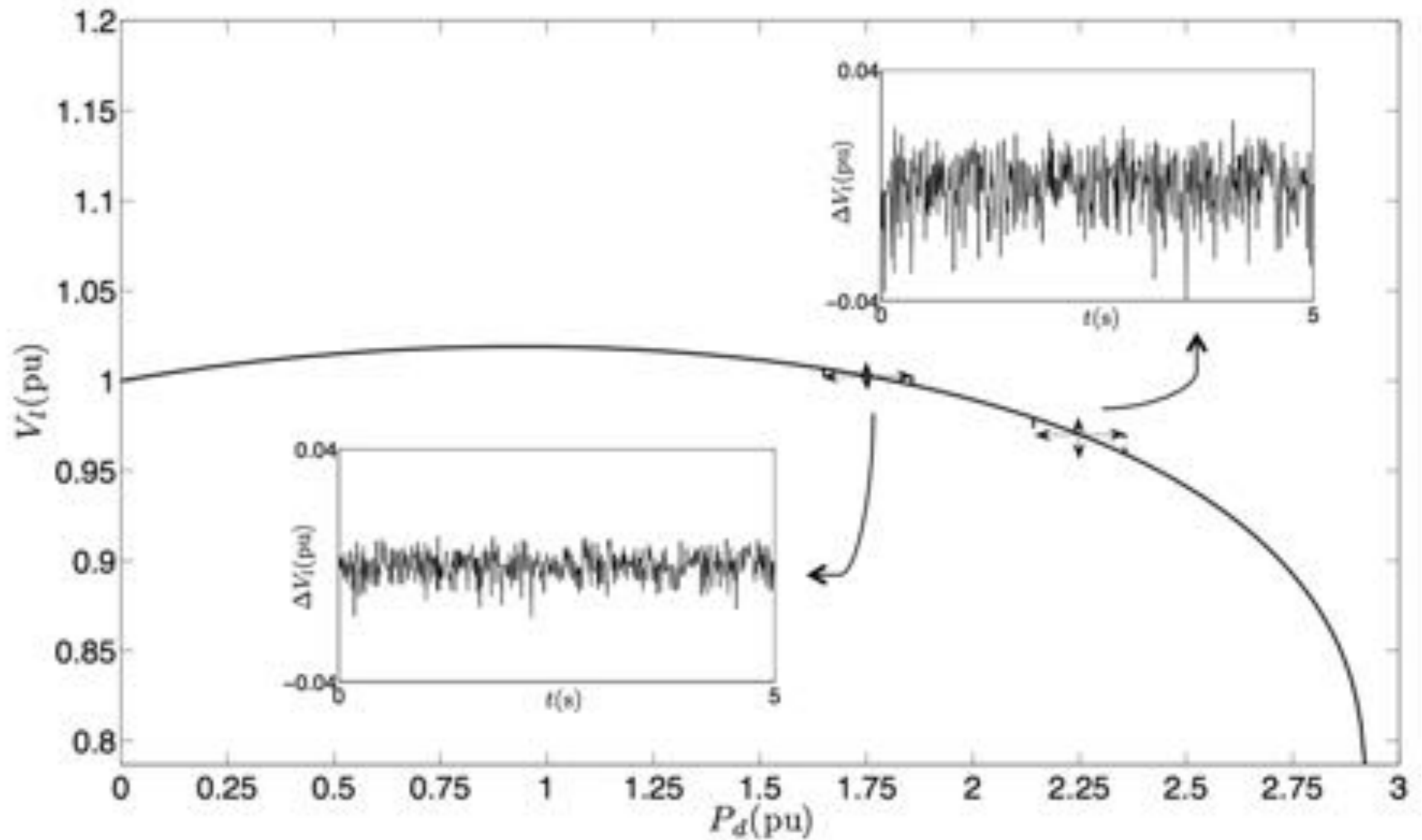


**Which statistics provide
useful (detectable)
early warning?**

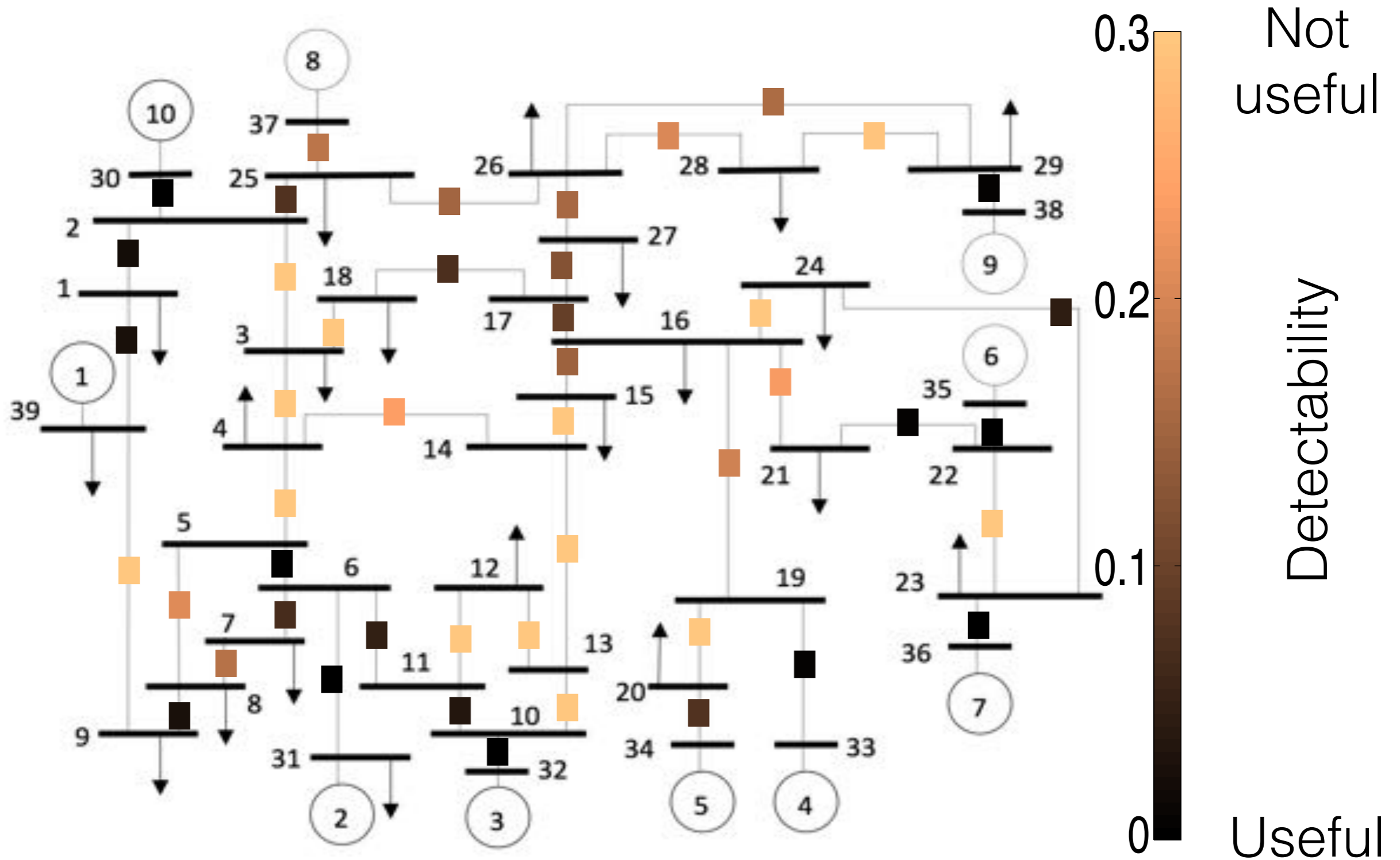
Variance of voltages



Why is variance in voltage useful?

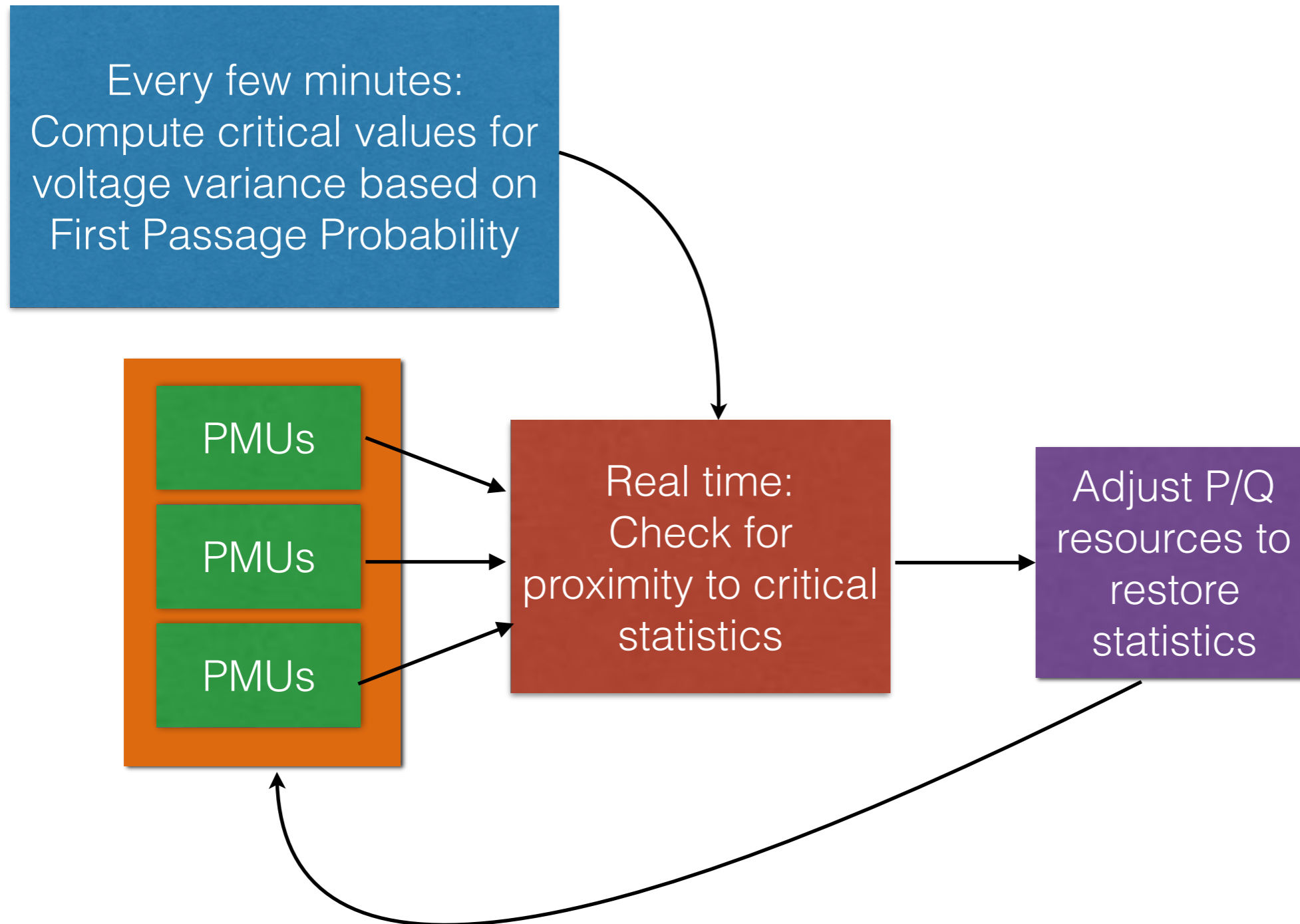


Autocorrelation of currents

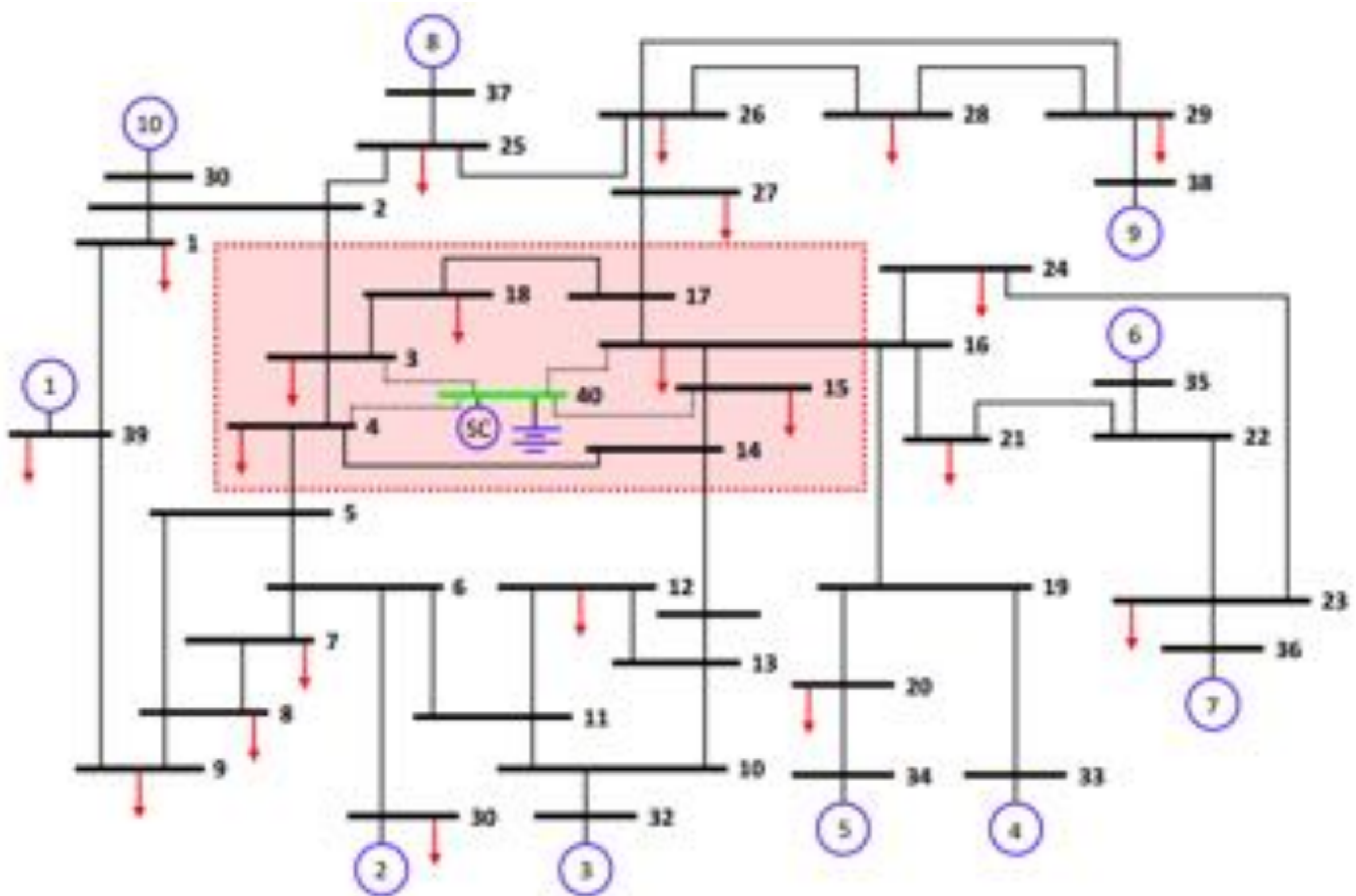


Can we use these signals
to build a control system?

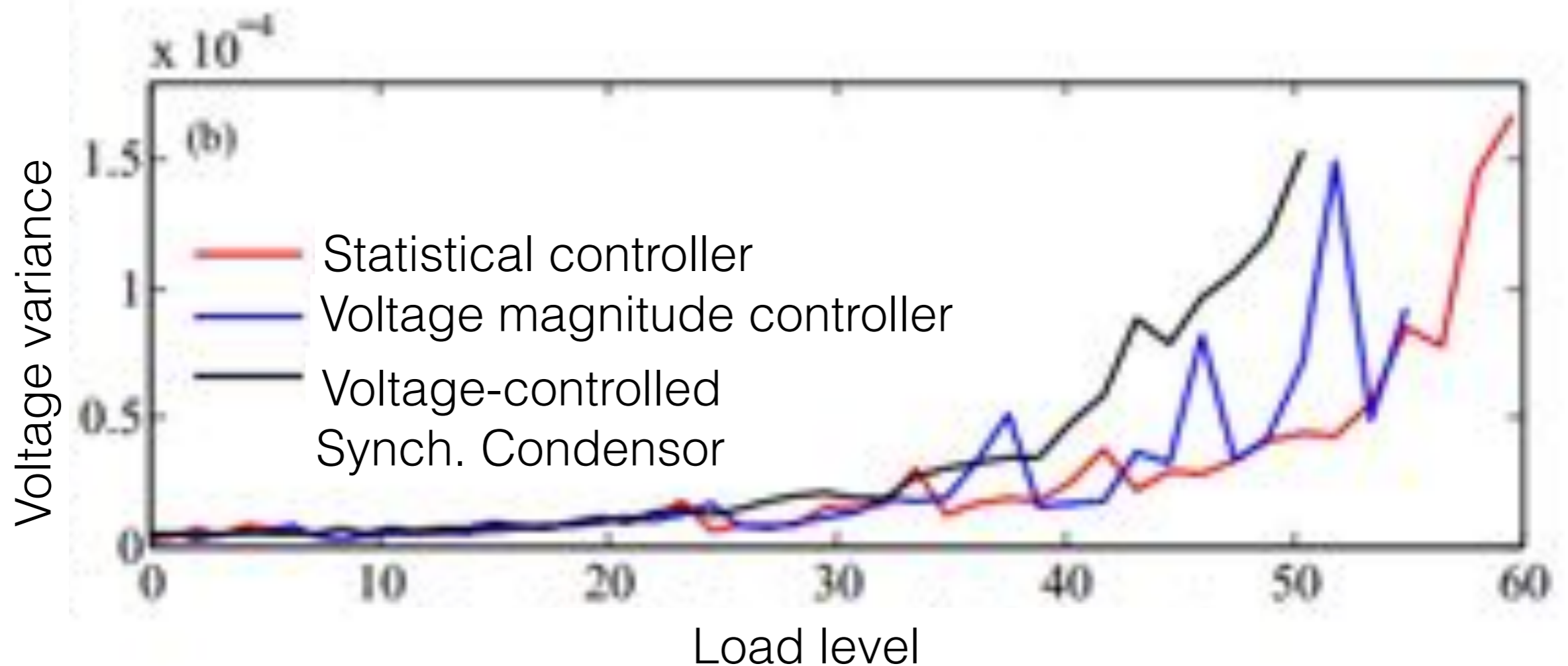
A (simple) control system



Apply to 39 bus test case



Statistical feedback allows us to make better control decisions relative to voltage-magnitude-based control



In summary

- **Autocorrelation** and **variance** are, *sometimes*, useful indicators of proximity to instability.
- Variances of **voltages near loads** are consistently good indicators of proximity to voltage collapse, even when voltage magnitudes are not.
- Autocorrelations of **currents near generators** (particularly smaller ones) are generally good indicators of system-wide stability issues (e.g., inter-area oscillations—Hopf bifurcation)
- These statistics can be used to design statistical control systems that can **improve voltage stability**.

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NY city, Nov. 9, 1965
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