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Probability Density Methods for the Analysis of Power Grids Under Uncertainty

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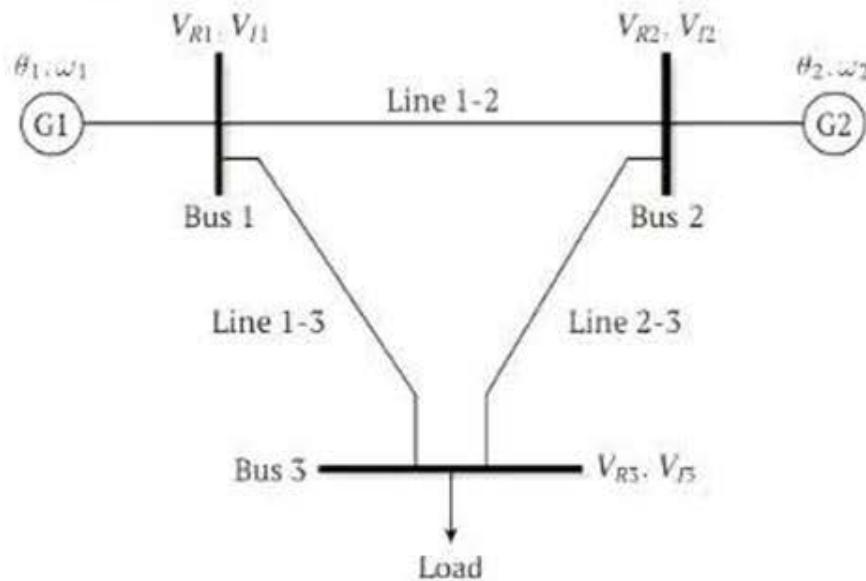
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SIAM Conference on Computational Science & Engineering 2015

Equations for generators of wind turbines



Index-1 DAE system for N machines, $i \in [1, N]$:

$$\frac{d}{dt} \theta_i(t) = \omega_B \Delta \omega_i,$$

$$2H_i \frac{d}{dt} \omega_i(t) = -D_i \Delta \omega_i + P_{e,i} + P_{m,i}(t).$$

$$0 = \mathbf{g}(\boldsymbol{\theta}, \mathbf{V})$$

θ_i	Phase angle
ω_i	Angular speed
$V_{R,i}, V_{I,i}$	Bus voltages
$P_{e,i}(\boldsymbol{\theta}, \mathbf{V})$	Electric power
D_i	Damping factor
H_i	Inertia
ω_B	Base angular speed

Mechanical power (uncertain):

$$P_{m,i}(t) = \langle P_{m,i}(t) \rangle + P'_{m,i}(t)$$

$$\langle P'_{m,i}(t) P'_{m,j}(s) \rangle = \sigma_{ij}^2 \rho_{ij}(t, s)$$

PDF method

Objective: Derive a PDE governing the joint PDF of phase angles and angular speeds, $p(\mathbf{x}, t)$, $\mathbf{x} = (\theta_1, \omega_1, \dots, \theta_N, \omega_N)^\top \in \mathbb{R}^{2N}$

Mathematical challenges:

1. $\rho_{ij}(t, s) \neq \delta(t - s)$ (colored noise)
2. Various timescales need to be considered
 - Relaxation timescale $\gamma_i^{-1} = 2H_i/D_i$
 - Correlation timescale λ_{ij}

For $\rho_{ij}(t, s) = q_{ij}\delta(t - s)$ (white noise),

$$\frac{\partial p}{\partial t} + \nabla \cdot (\mathbf{v}(\mathbf{x})p) = \sum_{i=1}^N D_{\omega_i \omega_i} \sum_{j=1}^N \frac{\partial^2 p}{\partial \omega_i \omega_j} = \nabla \cdot (\mathbf{D} \nabla p)$$

$\mathbf{v}(\mathbf{x})$: Drift velocity (deterministic part of DAE system)

$D_{\omega_i \omega_j} = \sigma_{ij}^2/2$: Components of diffusion tensor \mathbf{D}

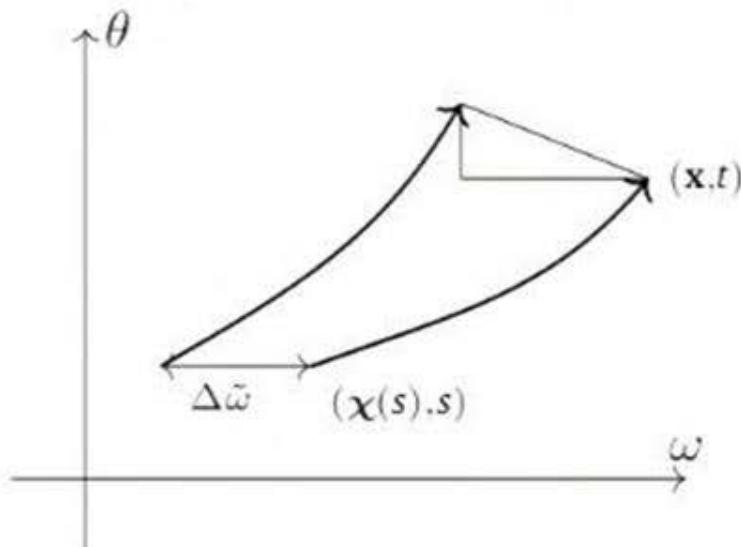
Classical LED closure

We employ a large-eddy diffusivity (LED) closure, $O(\sigma^2)$ accurate:

$$\frac{\partial p}{\partial t} + \nabla \cdot (\mathbf{v}p) = \sum_{i=1}^N \mathcal{D}_i[p]$$

$$\mathcal{D}_i[p(\mathbf{x}, t)] = \int_0^t \left| \frac{\partial \chi(s)}{\partial \mathbf{x}} \right| \langle P'_{m,i}(t) P'_{m,j}(s) \rangle \frac{\partial}{\partial \tilde{\omega}_j} p(\chi(s), s) \, ds,$$

where $\chi = (\tilde{\theta}_1, \tilde{\omega}_1, \dots, \tilde{\theta}_N, \tilde{\omega}_N)^\top$, $\dot{\chi}(s) = \mathbf{v}(\chi)$, $\chi(t) = \mathbf{x}$



Classical localization (Kraichnan):
Assume $\nabla_\chi p(\chi(s), s) \approx \nabla p(\mathbf{x}, t)$
for $|t - s| < \lambda_{ij}$, so that

$$\mathcal{D}_i[p] \approx \left(\int_0^t \langle P'_{m,i}(t) P'_{m,j}(s) \rangle \right) \frac{\partial}{\partial \tilde{\omega}_j} p$$



Classical localization results in the PDF equation

$$\frac{\partial p}{\partial t} + \nabla \cdot (\mathbf{v}(\mathbf{x})p) = \sum_{i=1}^N D_{\omega_i \omega_i}(t) \sum_{j=1}^N \frac{\partial^2 p}{\partial \omega_i \omega_j} = \nabla \cdot (\mathbf{D}(t) \nabla p)$$

For $\lambda_{ij}\gamma_i \gg 1$, the SDAEs exhibit time-scale separation

- ▶ ω_i : Fast variables, $\rightarrow \omega_S$ over the time scale $\gamma_i^{-1} \ll \lambda_{ij}$
- ▶ θ_i : Slow variables

The classical approximation does not recover such behavior



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Modified LED closure

Deconvolve by considering only the advective contribution to the evolution of p from s to t :

$$\left| \frac{\partial \chi(s)}{\partial \mathbf{x}} \right| p(\chi(s), s) = p(\mathbf{x}, t) \rightarrow \mathcal{D}_i[p] = \left(\int_0^t ds \langle P'_{m,i}(t) P'_{m,j}(s) \rangle \frac{\partial}{\partial \tilde{\omega}_j} \right) p(\mathbf{x}, t)$$

$$\mathcal{D}_i[p] = \left(\int_0^t ds \langle P'_{m,i}(t) P'_{m,j}(s) \rangle \sum_{k=1}^N \left\{ \frac{\partial \omega_k}{\partial \tilde{\omega}_j} \frac{\partial}{\partial \omega_k} + \frac{\partial \theta_k}{\partial \tilde{\omega}_j} \frac{\partial}{\partial \theta_k} \right\} \right) p(\mathbf{x}, t)$$

$$\frac{\partial p}{\partial t} + \nabla \cdot (\mathbf{v}(\mathbf{x})p) = \sum_{i=1}^N \frac{\partial}{\partial \omega_i} \left\{ \sum_{j=1}^N D_{\omega_i \omega_j}(\mathbf{x}, t) \frac{\partial p}{\partial \omega_j} + D_{\omega_i \theta_j}(\mathbf{x}, t) \frac{\partial p}{\partial \theta_j} \right\}.$$

The challenge is to estimate the DAE sensitivities $\partial \omega_k / \partial \tilde{\omega}_j$, $\partial \theta_k / \partial \tilde{\omega}_j$, i.e., components of the sensitivity matrix

$$\Psi(t | \chi(s), s) = \mathcal{T} \exp \left(\int_s^t \mathbf{J}(\chi(s'), s') ds' \right), \quad J_{ij}(\mathbf{x}, t) = \partial v_i(\mathbf{x}, t) / \partial x_j$$

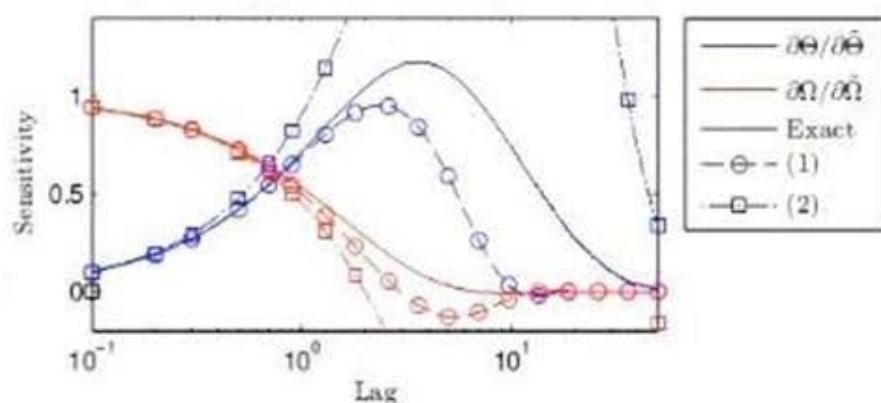
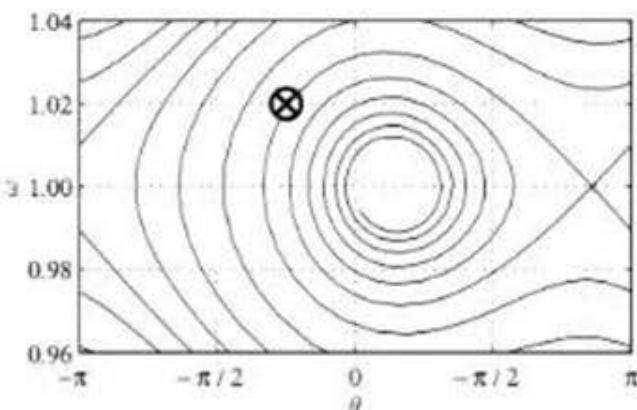
Diffusion coefficients

Various closed-form approximations to diffusion coefficients $D_{\omega_i \omega_j}$, $D_{\omega_i \theta_j}$.

1. $\lambda_{ij} \gamma_i > 1$: Linearize time-order exp. around (\mathbf{x}, t) :

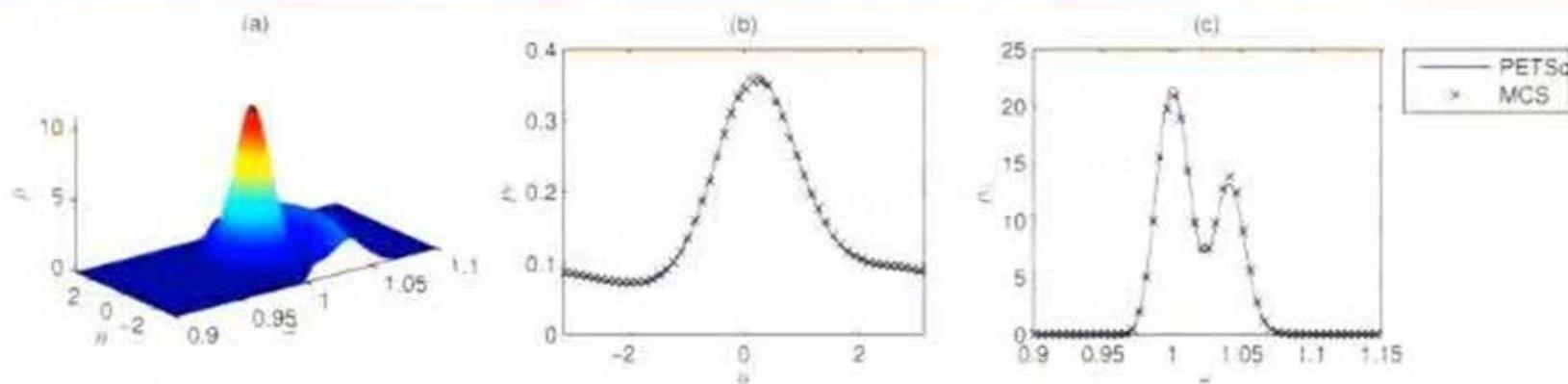
$$\Psi(t|\chi(s), s) = \mathcal{T} \exp \left(\int_s^t \mathbf{J}(\chi(s'), s') ds' \right) \approx \exp \{(t-s)\mathbf{J}(\mathbf{x}, t)\}$$

2. $\lambda_{ij} \gamma_i \ll 1$: Power series in λ_{ij} of time-ordered exponential

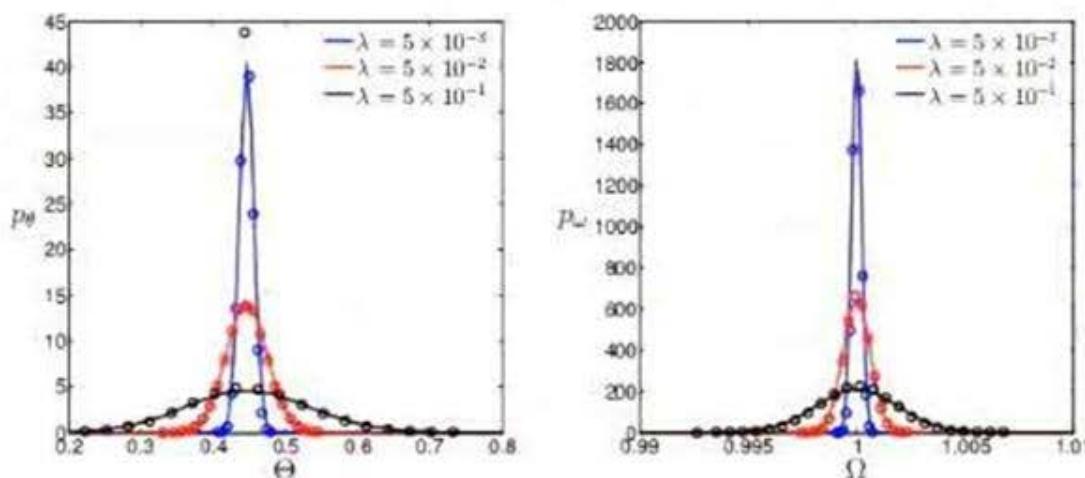


Comparison of various approximations to diffusion coefficients, $\lambda\gamma = 5$

Numerical experiments

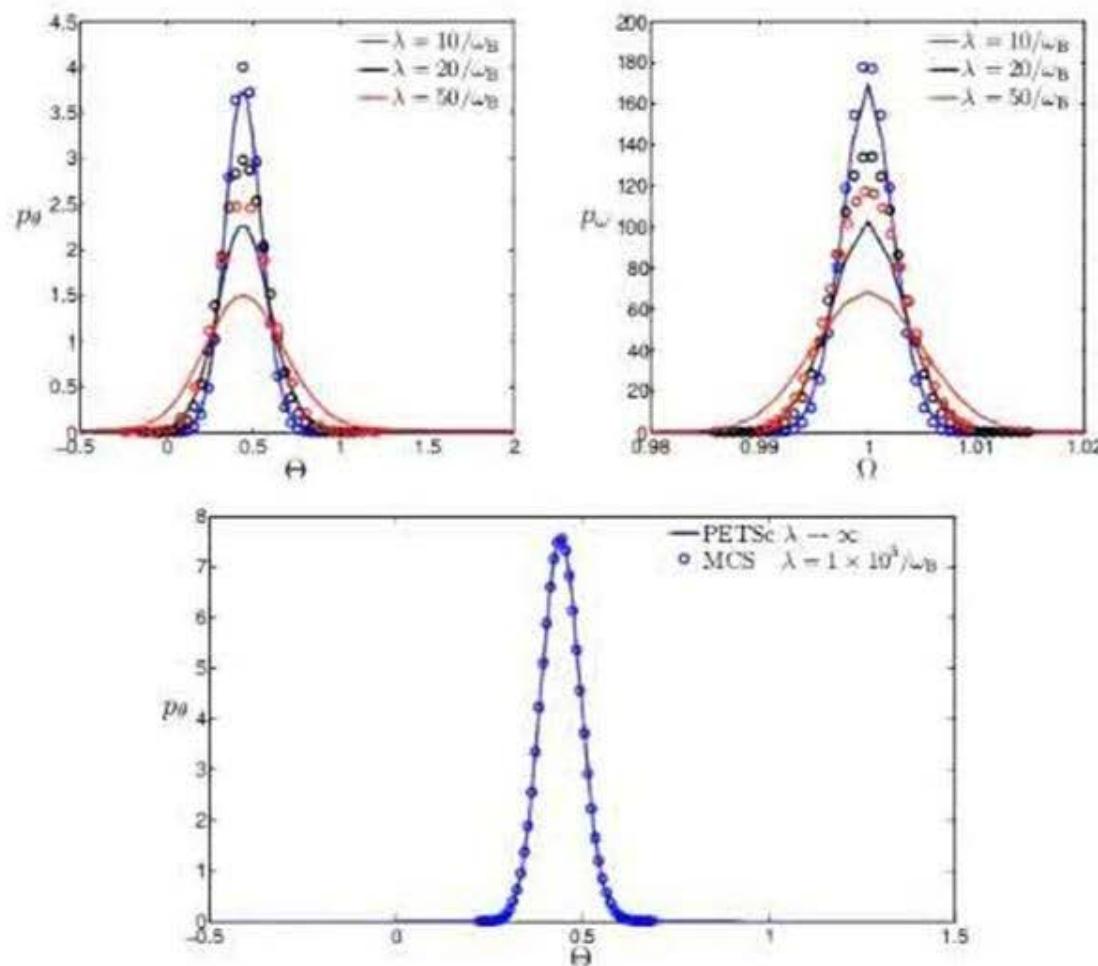


(a) Joint PDF of SMIB system. (b)–(c) Marginal PDFs of θ and ω



Marginal PDF snapshots

Numerical experiments



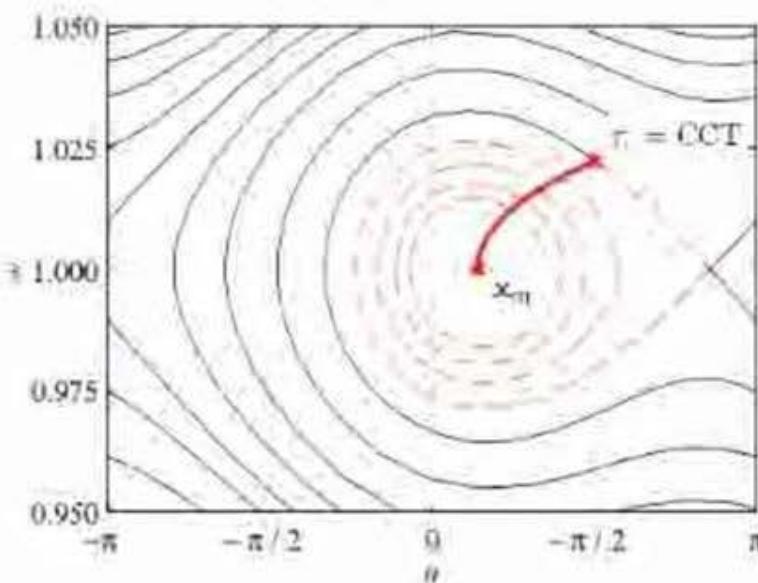
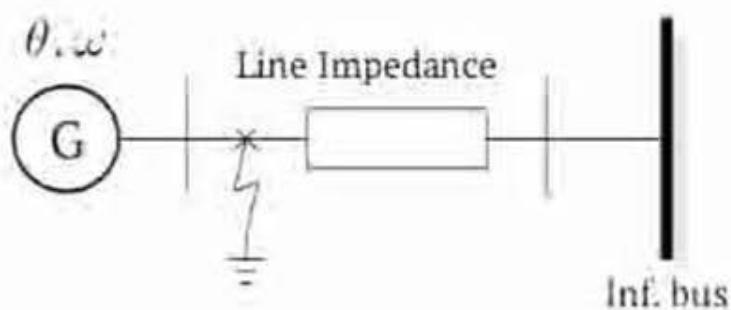
Stationary marginal PDFs

Transient stability under uncertainty

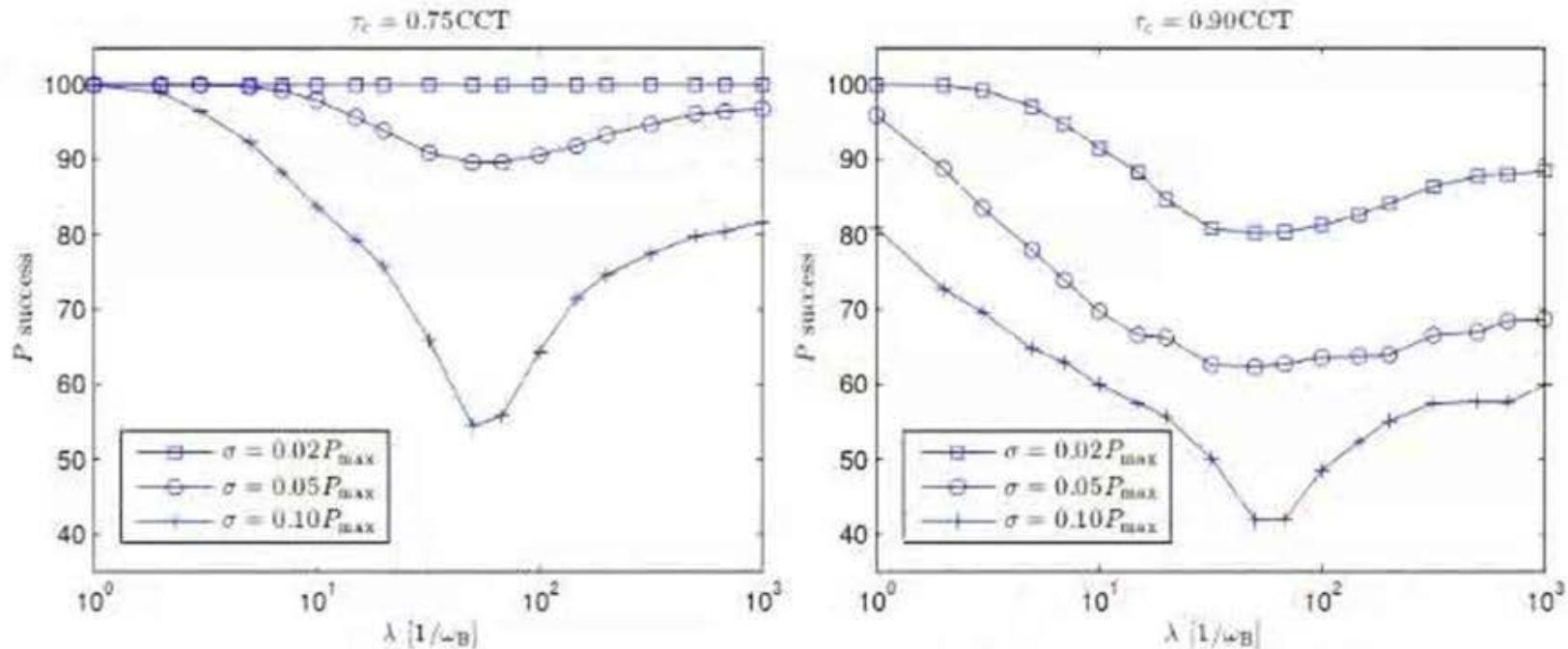
Failure: Catastrophic, unrecoverable event (i.e., relay triggering after fault clearing)

Due to uncertainty in P_m , a power system has non-trivial probability of failure after clearing a fault in sub-critical time τ_c

Example: For SMIB, define failure as $|\omega - \omega_s| \geq 0.05\omega_s$



Transient stability under uncertainty



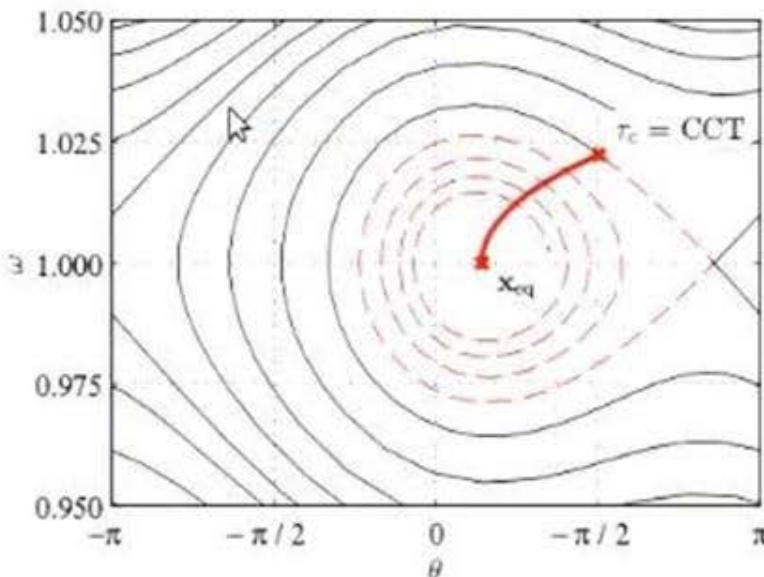
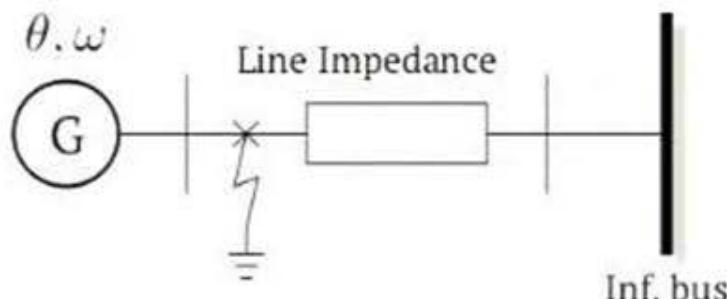
Probability of maintaining stability after clearing of fault, $\tau_c < \text{CCT}$

Transient stability under uncertainty

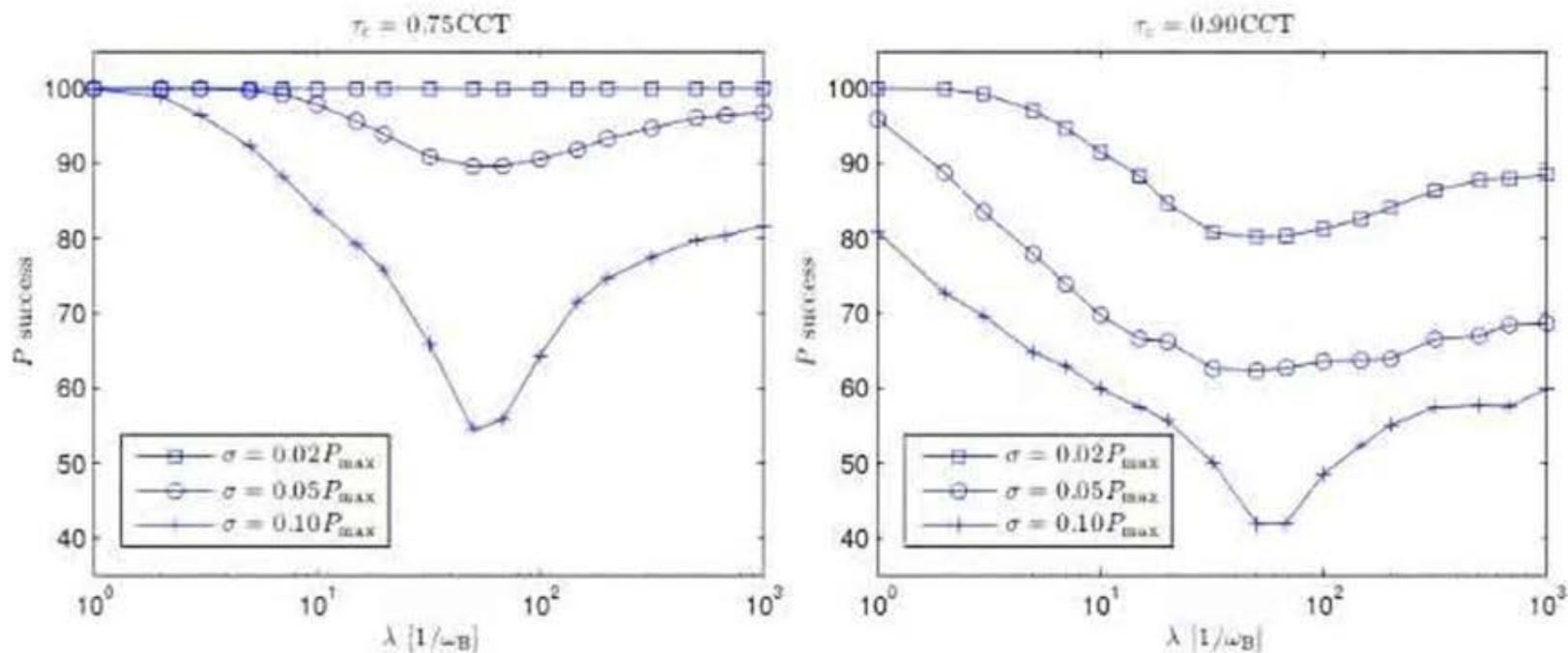
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