

# Probabilistic Small-Disturbance Stability Analysis of Renewable-rich Power Systems

**Kazi Hasan**, RMIT University, Australia

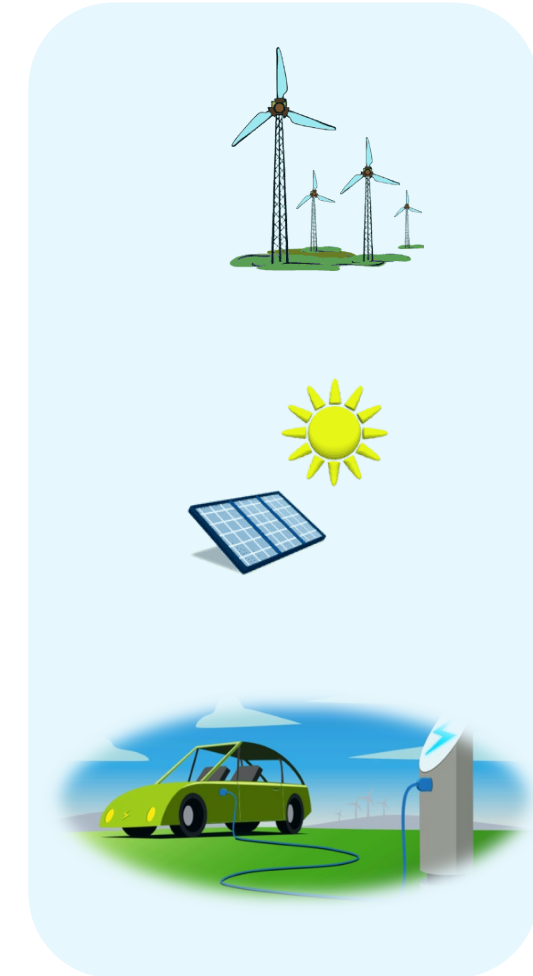
Acknowledgement: **Dr. Robin Preece**, **Prof. Jovica Milanović**, The University of Manchester, UK

# Outline of the Presentation

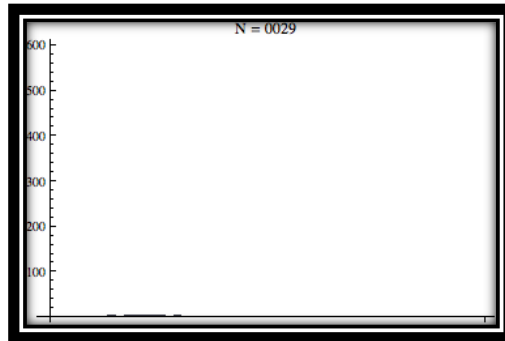
- Motivation for Probabilistic Stability Studies
- Uncertainty Modelling and Probabilistic Analysis Framework
  - Input Modelling
  - Computational Techniques
  - Applications to Power Systems
- Conclusions
  - Accuracy vs. Efficiency

# Motivation for Probabilistic Stability Analysis

- **Operational uncertainty** is increasing in power systems due to the integration of new energy resources
  - As a short-term phenomenon, **system stability is significantly affected** by the intermittent **generation** and behind-the-meter **demand** variability
  - **Probabilistic analysis** is highly required
- 
- KN Hasan, R Preece, JV Milanović, Priority ranking of critical uncertainties affecting small-disturbance stability using sensitivity analysis techniques, *IEEE Transactions on Power Systems*, 32 (4), 2016.
  - KN Hasan, R Preece, JV Milanović, The Influence of Load on Risk-based Small-Disturbance Security Profile of a Power System, *IEEE Transactions on Power Systems*, 33 (1), 2018.
  - KN Hasan, R Preece, Influence of Stochastic Dependence on Small-Disturbance Stability and Ranking Uncertainties, *IEEE Transactions on Power Systems*, 33 (3), 2018.
  - B Qi, KN Hasan, JV Milanović, Identification of Critical Parameters Affecting Voltage and Angular Stability Considering Load-Renewable Generation Correlations, *IEEE Transactions on Power Systems*, 34(4), 2019.
  - KN Hasan, R Preece, JV Milanović, Existing approaches and trends in uncertainty modelling and probabilistic stability analysis of power systems with renewable generation, *Renewable and Sustainable Energy Reviews*, vol 101, 2019.
  - RF Mochamad, R Preece, KN Hasan, Probabilistic multi-stability operational boundaries in power systems with high penetration of power electronics, *International Journal of Electrical Power & Energy Systems*, 135, 107382, 2022.

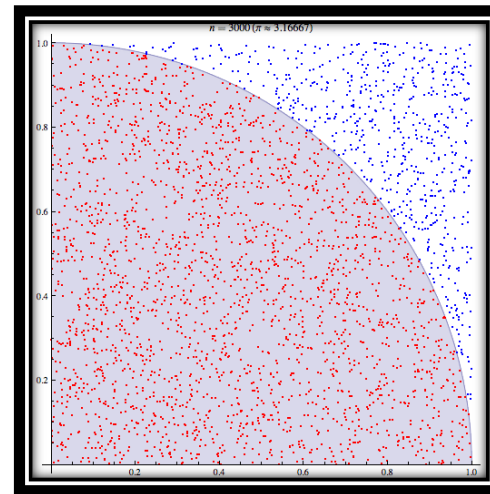


# Framework for Probabilistic Stability Analysis



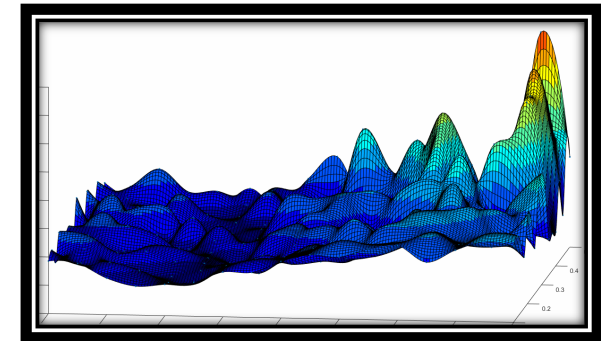
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Input parameter modelling



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Computational methods

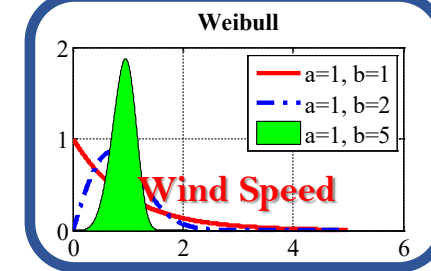
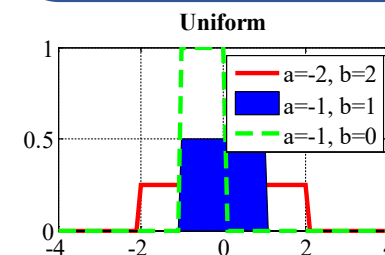
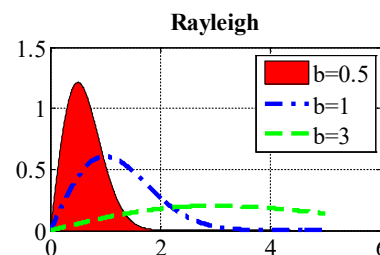
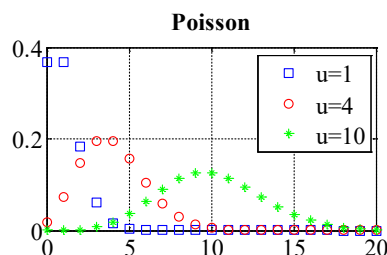
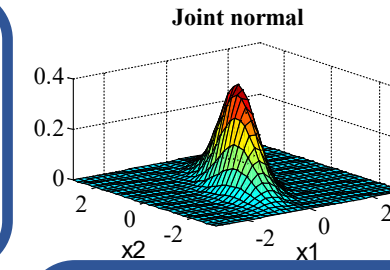
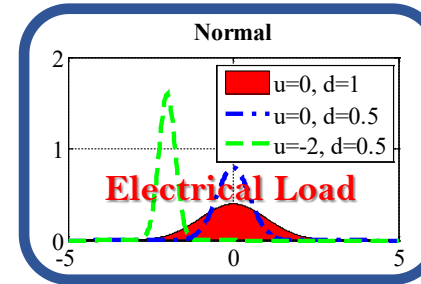
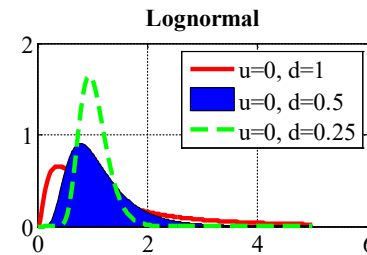
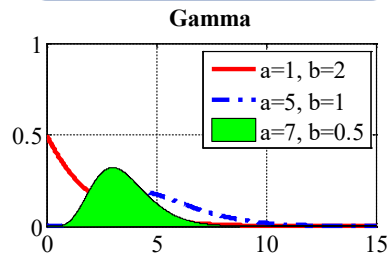
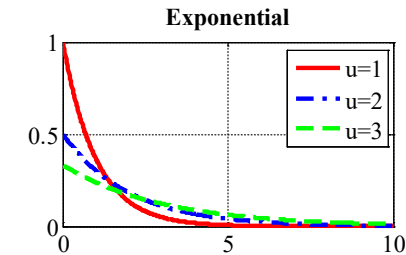
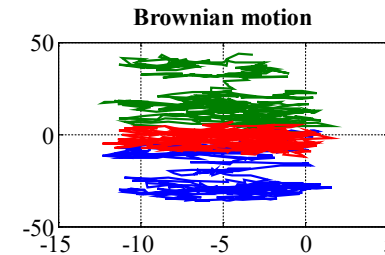
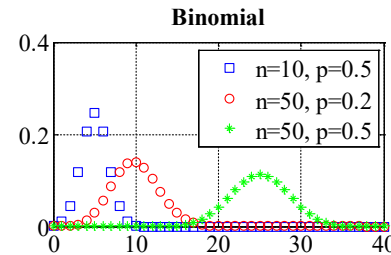
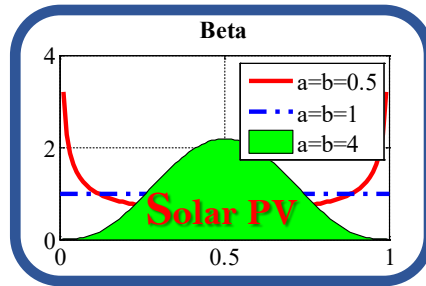


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Defining output indices

- KN Hasan, R Preece, JV Milanović, Existing approaches and trends in uncertainty modelling and probabilistic stability analysis of power systems with renewable generation, *Renewable and Sustainable Energy Reviews*, vol 101, 2019.

# 1 Probability Distribution of Power System Input Parameters



# 1 Probability Distribution of Power System Input Parameters

Variables	Independent Probability distributions
<b>Operational variables</b>	
Power system load	Normal [19-40], Gumbel [41, 42], discrete Normal [43], pdf of past data [44, 45], joint Normal [46-48], Brownian motion [49]
Wind speed/power	Weibull [20, 21, 24, 27-34, 36, 41, 50-53], Normal [22, 35], discrete Normal [43, 54], joint Gaussian [55, 56], log-normal [52, 53, 57], gamma [52]
Solar power	Beta [27-36], Weibull [21, 58], pdf of past data [44]
Power generation	Normal [20, 23], historical system data and statistics [47]
EV demand	Normal [15, 20]
EV charging time	Exponential [15, 20]
EV customer arrival	Poisson [20]
Generator fuel price	Geometric Brownian motion [21]
<b>Disturbance variables</b>	
Fault clearing time	Normal [23, 41, 48, 59-61], Poisson jump process [62]
Fault type	Historical system data and statistics [23, 41, 45, 59], pdf of past data [45]
Fault location	Historical data [41, 60], uniform [15, 23, 59], empirical pdf of data [45]
Fault incident	Poisson [59, 63, 64], binomial [64]
Fault impedance	Normal [23, 41]
Fault duration	Rayleigh [23, 41]
Load change	Poisson jump process [15, 62]
Contingency/failure	Markov model [43, 46]
Transformer failure	Arrhenius-Weibull [65], Normal [65]
Time to failure	Exponential [63]
Unsuccessful reclose	Bernoulli [62]
HVDC failure	Weibull [66]
Transformer overloading	Binomial [61, 67]

KN Hasan, R Preece, JV Milanović, Existing approaches and trends in uncertainty modelling and probabilistic stability analysis of power systems with renewable generation, *Renewable and Sustainable Energy Reviews*, vol 101, 2019.

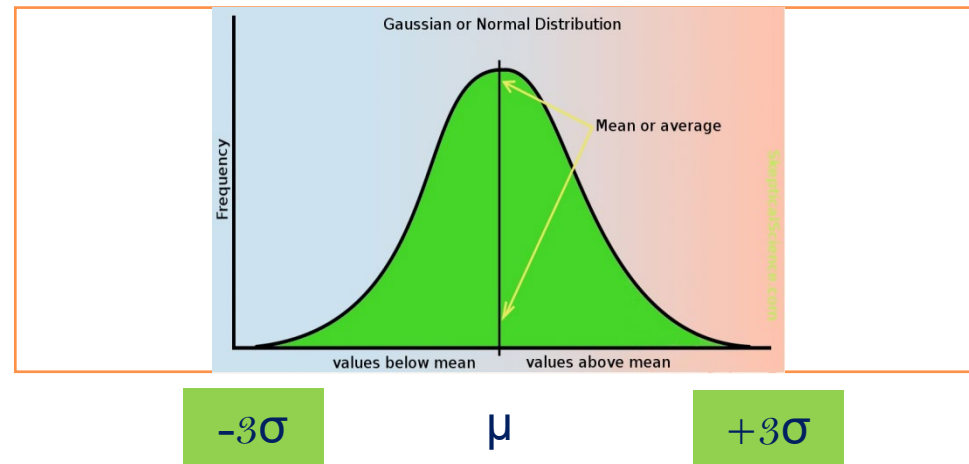
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# Probabilistic Computational Methods

## Monte Carlo Simulation

The MC simulation involves,

- Defining a domain of possible inputs,
- Generating inputs randomly from a probability distribution over the domain,
- Performing deterministic computations on the inputs, and
- Aggregating the results.



# Probabilistic Computational Methods

Monte Carlo Simulation is **accurate** but **computationally expensive**

The required number of Monte Carlo simulation can be calculated by using the stopping rule formula presented below.

$$E > \left[ \left\{ \Phi^{-1}(1 - \delta/2) \cdot \sqrt{\sigma^2(X)/N} \right\} / \bar{X} \right]$$

In (1),  $\Phi^{-1}(\bullet)$  represents the inverse Normal *cdf* with a mean of zero and standard deviation of one,  $\sigma^2(\bullet)$  is the variance of a sample, and  $\delta$  is the desired confidence level. As presented in (1), simulations can be stopped if the calculated sample mean error falls below a specific threshold,  $E$ . This is true assuming that the outputs follow Normal distribution (which might not be true in many cases).

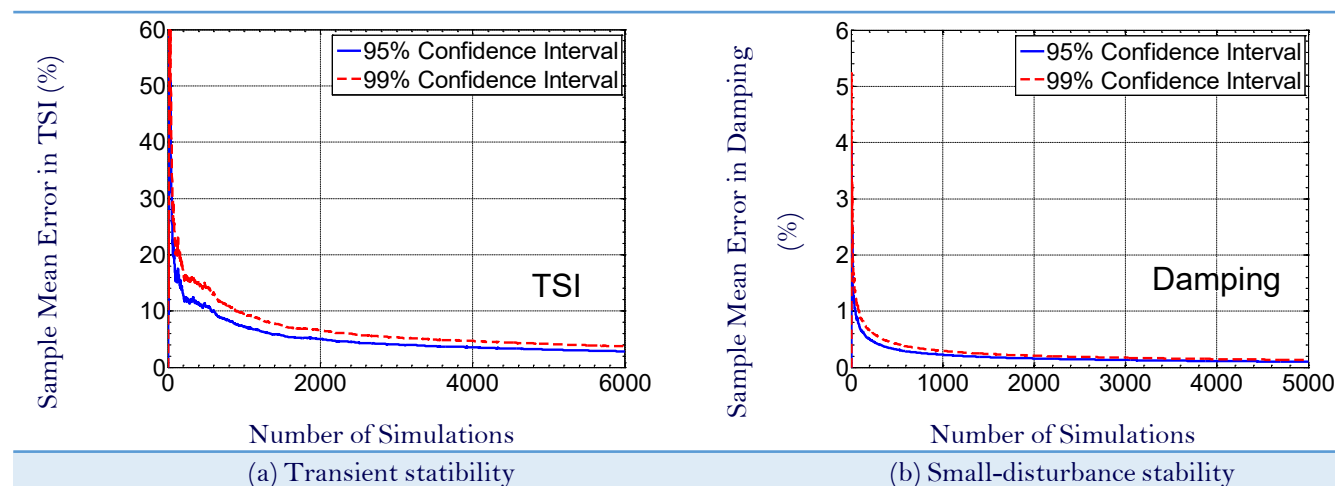


Figure: Illustrative results of Monte Carlo stopping rule.

- KN Hasan, R Preece, JV Milanović, Existing approaches and trends in uncertainty modelling and probabilistic stability analysis of power systems with renewable generation, *Renewable and Sustainable Energy Reviews*, vol 101, 2019.



## 2 Probabilistic Computational Methods

Table: Probabilistic techniques and stability applications.

Amount of research interest of different computational methods with different stability studies				
	Transient Stability	Small-disturbance Stability	Voltage Stability	Frequency Stability
Monte Carlo (MC)	High	High	High	High
Sequential MC	Medium	None	None	Low
Quasi-MC	None	Low	None	None
Markov Chain MC	Low	Low	None	None
Point Estimate (PEM)	Low	Medium	Low	None
Cumulant-based	None	Low	Low	None
Probabilistic Collocation	Low	Medium	Low	None

- KN Hasan, R Preece, JV Milanović, Existing approaches and trends in uncertainty modelling and probabilistic stability analysis of power systems with renewable generation, *Renewable and Sustainable Energy Reviews*, vol 101, 2019.

## 3 Probabilistic Outputs

OUTPUT DISTRIBUTION IS AFFECTED BY THE INPUT UNCERTAINTIES

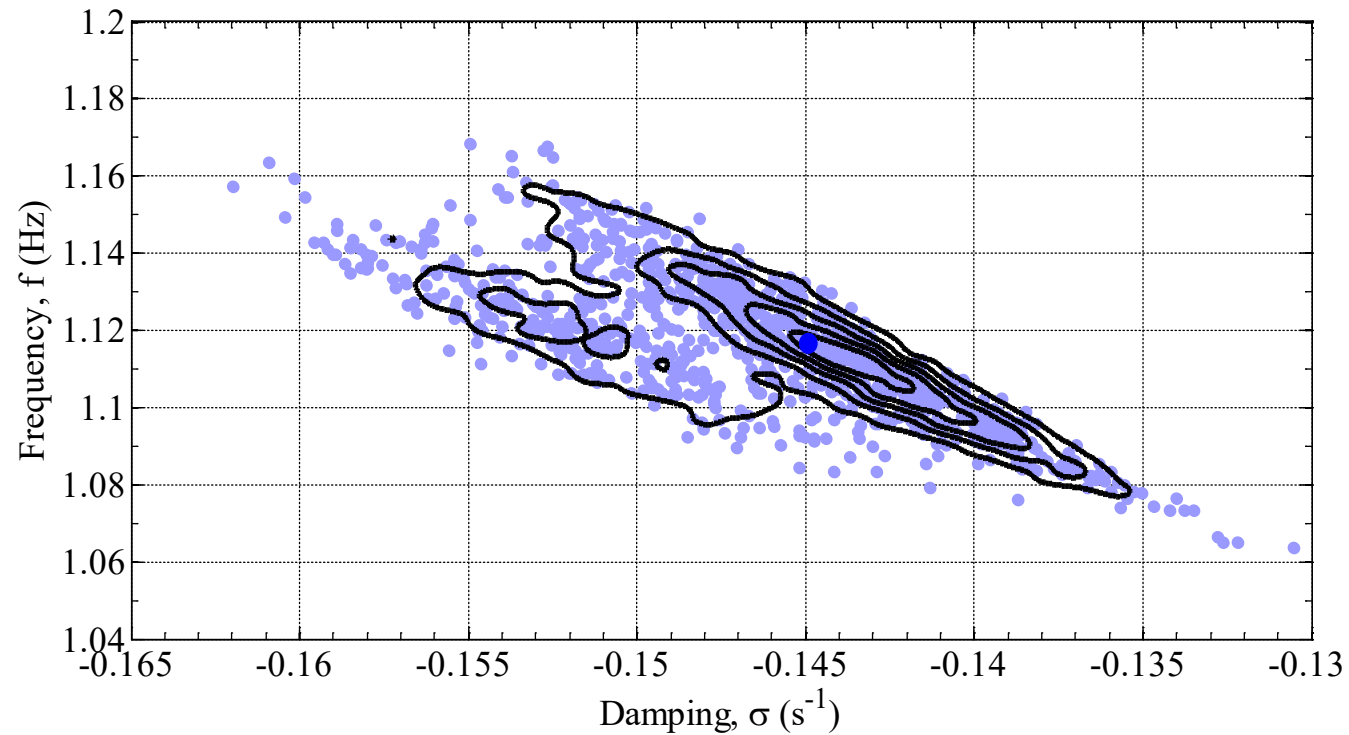


Figure: Contour plot and footprints of critical eigenvalue on the complex plane as affected by input parameter uncertainties.

# Next Phases of the Work ... (1/5)

Application of **sensitivity analysis** techniques

IEEE TRANSACTIONS ON POWER SYSTEMS, VOL. 32, NO. 4, JULY 2017

2629

## Priority Ranking of Critical Uncertainties Affecting Small-Disturbance Stability Using Sensitivity Analysis Techniques


Kazi Nazmul Hasan, *Member, IEEE*, Robin Preece, *Member, IEEE*, and Jovica V. Milanović, *Fellow, IEEE*

# Next Phases of the Work ... (2/5)

Application of **game theory** approaches

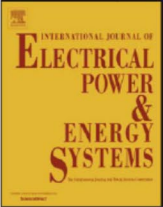
Electrical Power and Energy Systems 97 (2018) 344–352

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Application of game theoretic approaches for identification of critical parameters affecting power system small-disturbance stability

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# Next Phases of the Work ... (3/5)

## Application of risk analysis

IEEE TRANSACTIONS ON POWER SYSTEMS, VOL. 33, NO. 1, JANUARY 2018

557

### The Influence of Load on Risk-Based Small-Disturbance Security Profile of a Power System

Kazi N. Hasan , *Member, IEEE*, Robin Preece , *Member, IEEE*, and Jovica V. Milanović , *Fellow, IEEE*



# Next Phases of the Work ... (4/5)

Application of **correlation modelling**

IEEE TRANSACTIONS ON POWER SYSTEMS, VOL. 33, NO. 3, MAY 2018

3227

## Influence of Stochastic Dependence on Small-Disturbance Stability and Ranking Uncertainties


Kazi N. Hasan , *Member, IEEE* and Robin Preece , *Member, IEEE*

# Next Phases of the Work ... (5/5)

Application of **multi-stability** operational boundary identification

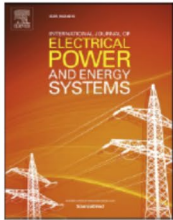
Electrical Power and Energy Systems 135 (2022) 107382


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**International Journal of Electrical Power and Energy Systems**

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Probabilistic multi-stability operational boundaries in power systems with high penetration of power electronics

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# Summary

- Operational uncertainties are increasing
- Probabilistic analysis is required
- Efficient computational techniques needs to be applied
- A probabilistic framework may be applied to further analysis:
  - Sensitivity analysis
  - Game theory application
  - Risk assessment
  - Correlation modelling





Thank you