Complex Systems, Complex Networks, and their Reliability

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- Objects:
- Phenomena:
- Methods and Tools:

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 - * Structural systems (tall buildings, bridges)
 - Complex Networks
 - * Technological networks (road networks, water distribution networks)
 - ★ Financial networks (stock market networks, banking networks)
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Notation:

- \bullet $x \in \mathbb{R}^d$ represents the uncertain excitation of a system
 - x is a random vector with joint PDF $\pi(x)$
- ullet $F\subset\mathbb{R}^d$ is a failure domain (unacceptable system performance)

$$F = \{x : g(x) \ge b^*\}$$

- g(x) is a performance function (loss function)
- b^* is a critical threshold for performance
- $I_F(x) = 1$ if $x \in F$ and $I_F(x) = 0$ if $x \notin F$



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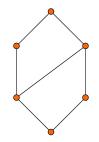
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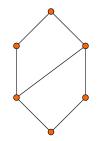
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<u>Idea:</u> To use <u>advanced simulation methods</u> e.g. Subset Simulation

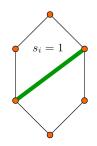
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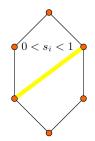
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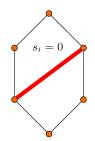
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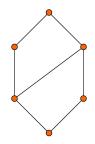
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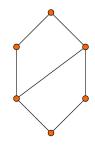
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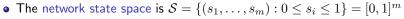
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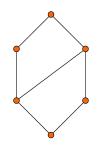
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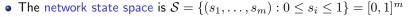
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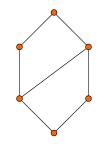
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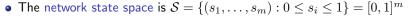
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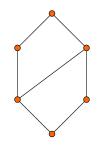
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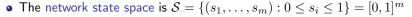
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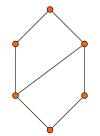
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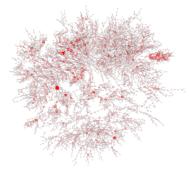
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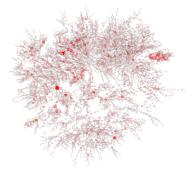


US Western States Power Grid, m = 6,594

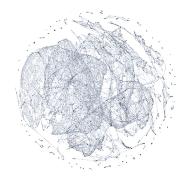
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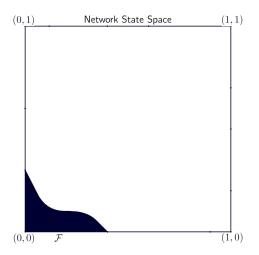
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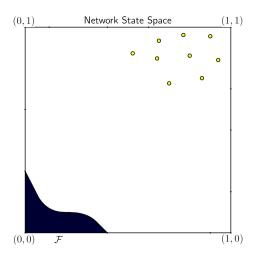
<u>Idea:</u> To use <u>efficient methods</u> developed by the "systems research community" e.g. Subset Simulation

Schematic Illustration

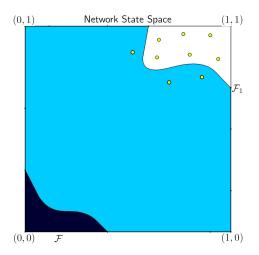


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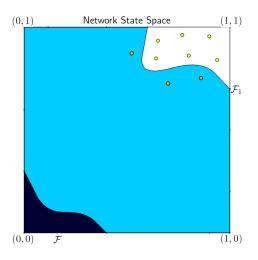




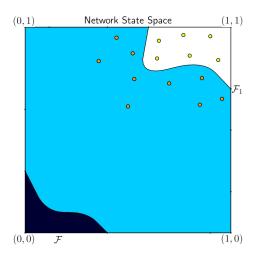
Monte Carlo Samples



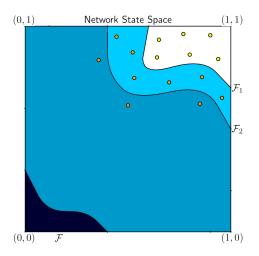
$$\mathbb{P}(\mathcal{F}_1) \approx \frac{1}{N} \sum_{i=1}^{N} I_{\mathcal{F}_1}(x_0^{(i)}) = p$$



"seeds"

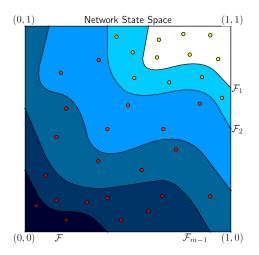


MCMC samples

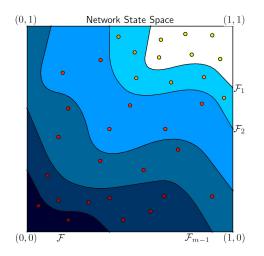


$$\mathbb{P}(\mathcal{F}_2|\mathcal{F}_1) \approx \frac{1}{N} \sum_{i=1}^{N} I_{\mathcal{F}_2}(x_1^{(i)}) = p$$





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$${\rm SS \ estimator}: \quad \ \widehat{p_F} = p^{m-1} \frac{N_F}{N}$$

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