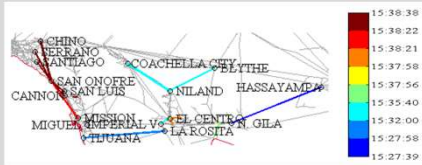
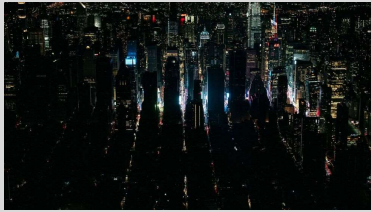


# Research 1: Failure Cascade Prediction in Power Grids

## Motivation

Large Blackout due to Failure Cascade

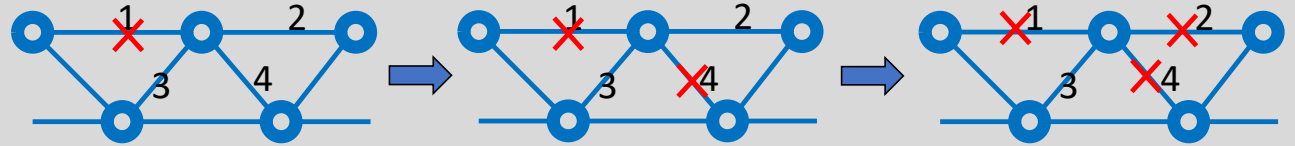


Q: Can we **predict** it in early stage?

## Contributions

- 1) **Influence model** to capture failure cascade
- 2) 3-step **learning framework** to learn the influence model
- 3) **Evaluation on real systems** over: failure size, final state, failure time, time cost, etc.
- 4) **Critical component identification**

## Influence Model



Pair-wise influence:  $A_{14} = \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix}$

Influence Weight:  $D_{12}, D_{42}$

Modeling failure cascade:

[ Parameters:  $\{A, D, \vec{\epsilon}\}$  ]

$$\tilde{s}_j[t+1] = \sum_{i=1}^N D_{ij} (A_{ij}^{(11)} s_i[t] + A_{ij}^{(21)} (1 - s_i[t]))$$

$$\hat{s}_j[t+1] = 1\{\tilde{s}_j[t+1] \geq \epsilon_j\}$$

## Learning Framework

$A$ : Monte-Carlo

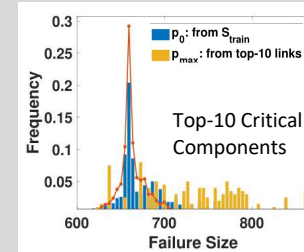
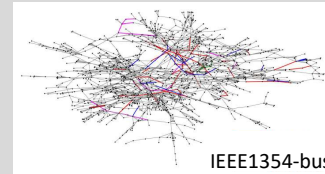
$D$ : Min. MSE  $\|s[t] - \tilde{s}[t]\|^2$

$\vec{\epsilon}$ : Nearest Neighbor

Advantage:

- Fully **parallelizable**
- $A$  and  $D$  **interpretable**

## Experiments



Metric	Result
Failure Size	<3% error
Final State	<9% error
Failure Time	<1.5 units
Time Cost	1/872 of flow cal.

## Connection to Industry Research:

- 1) Prediction framework is promising to failure cascade in server communication networks.
- 2) ML experience in time series prediction over large datasets