# Modeling Impact of Communication Network Failures on Power Grid Reliability

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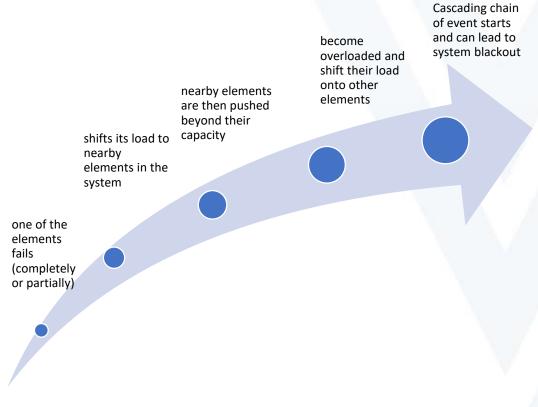
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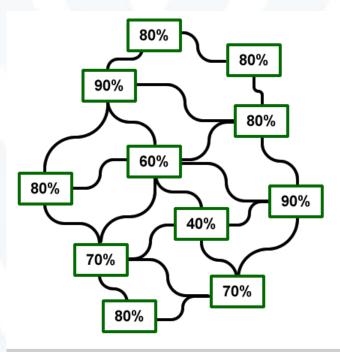




## **Cascading Failure in Power Grid**



balancing generation and load is critical



**Network running normally** 

Source: Wikipedia

### Why do cascading failures and blackouts occur?

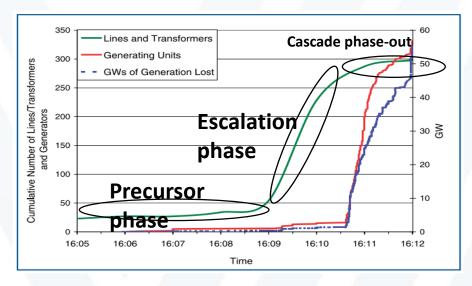
- Large blackouts result from the cascade of component failures in the transmission grid triggered by initial disturbances:
  - Natural disasters and human-related events such as unintentional human faults, sabotage occurrences and WMD attack.
  - Cascading failure exhibit three phases where during escalation phase high number of transmission lines fails in a small time window

#### Example: 2003 Northeast Blackout:

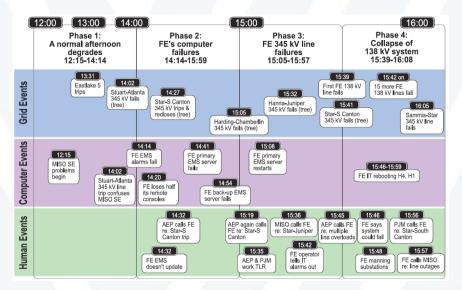
- Occurred due to a combination of transmission-line failure and communication network failure
- Alarm software failed leaving the human operators unaware of the transmission-line outage which contributed the cascading failure [1]

#### • Example: 2003 Italy Blackout:

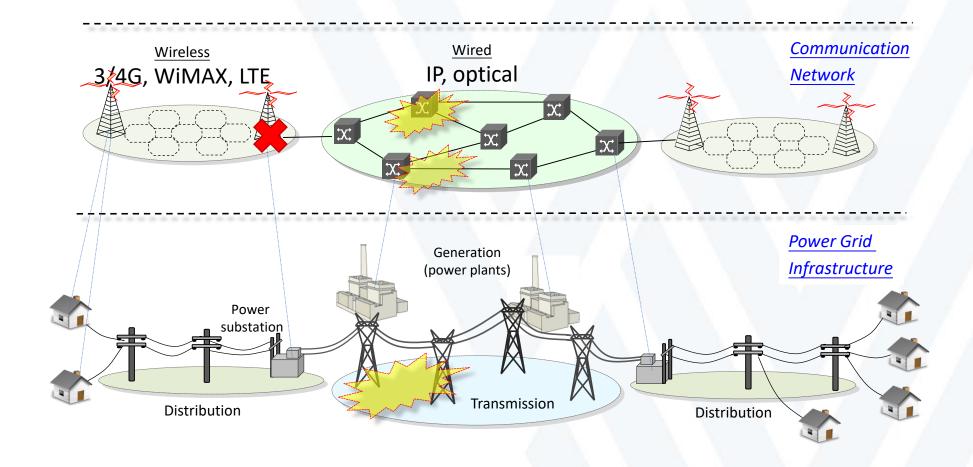
 During Power blackout at Italy, an unplanned power shutdown eventually led failures in the communication network, which in turn initiated a series of cascading failures in the power grid [2].



Time evolution of 2003 blackout in US and Canada exhibit three phases. [Online]. [3]



## Modeling requires a multi-layer view of the electric infrastructure



#### **Earlier work**

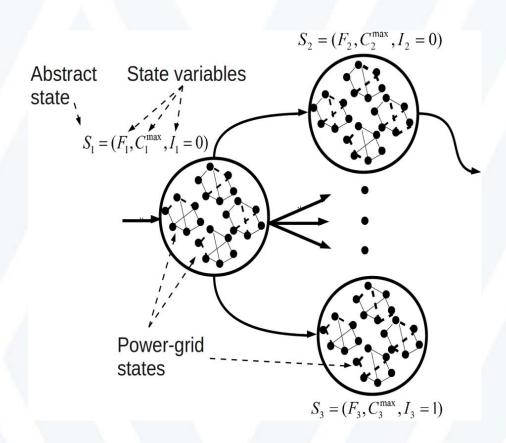
We used the following two models to define interdependency between power grid and communication network:

- Stochastic abstract-state evolution (SASE) [5] model :
  - Describes the dynamics of cascading failures based upon Markov chains
- Interdependent Markov-chain (IDMC) [4] model:
- A minimal MC that encompasses the individual MC for each physical system and their interdependencies

#### Review of SASE model

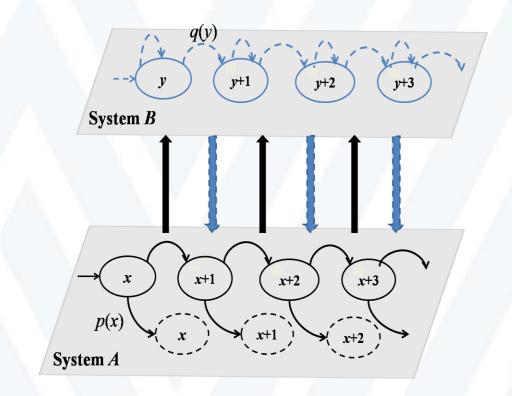
- Main ideas of the stochastic abstract-state evolution (SASE) approach:
  - Simplify the state space of the complex power system (equivalence classes)
  - Capturing the effects of the omitted variables through the transition probabilities and their parametric dependence on physical attributes and operating characteristics of the system.

- Aggregate state variables:  $S_i = (F_i, C_i^{\text{max}}, I_i)$ 
  - **F**: number of failed lines
  - **C**<sub>max</sub>: maximum capacity of failed lines
  - I: Cascade-stability of power grid



#### Review of Interdependent Markov-chain (IDMC) model

- Each network is represented by a Markov chain
  - Number of failures in the power grid: x
  - Number of failures in the communication system: y
- Failure in one chain is correlated with failure in the other chain via state-dependent coupling variables
- Transition probabilities are influenced by communication-network topology via state-dependent variables representing significance of failed nodes/links

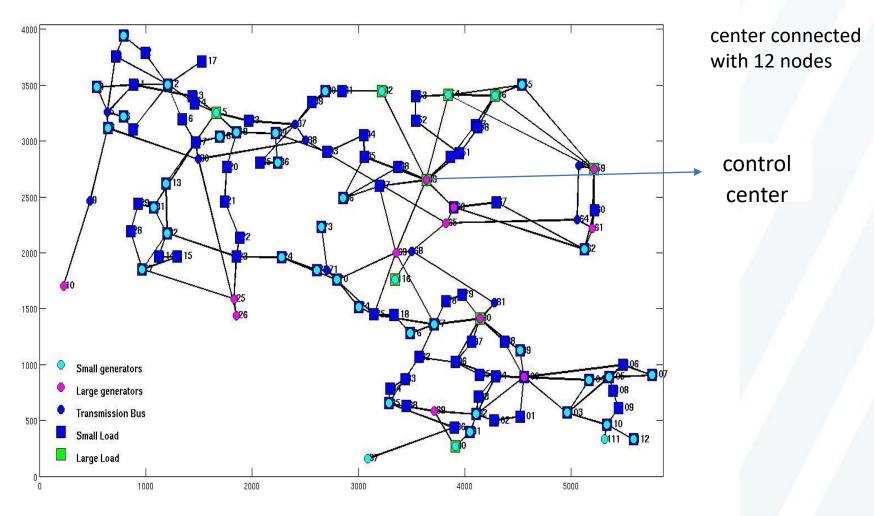


Inter-dependent Markov chains (IDMC) refers a coupling parameter to characterize the influence of communication network in power grid but simply considered it as a constant

## Proposed model to capture the impact of interdependency between power grid and communication network

- We introduce the coupling parameter d in terms of the minimum hop distance and the maximum node degree
  of the failed communication nodes.
- Our observations illustrate that, a decrement in the minimum hop distance or an increment in the maximum node degree of the failed communication nodes increase the cascading-failure probability in the power grid.
- By characterizing the coupling parameter *d*, we study the communication topology and identify it's impact on cascading-failures in power grid in an interdependent system environment.

### Communication/control network over-layed on IEEE-118 bus topology



- We selected the 49<sup>th</sup> node as the center node. Its node degree is 12, which is the highest
- Maximum hop distance from the center node is 8



#### Coupling between communication/control and transmission networks

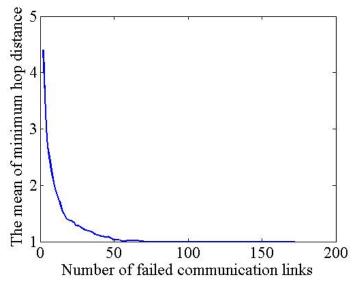
- Consequence of power loss on communication:
  - A failure in a transmission line triggers a communication-link failure in communication system with probability q.
- Consequence of communication-link failure on power loss:
  - Without communication influence, cascading failures stop in the power grid with probability p(x), which depends on the number of failures in the power grid: this is the cascade-stop probability. [earlier SASE model]
  - A communication-link failure reduces the cascade-stop probability in the power grid from p(x) to p(x)(1-d(y)), where d(y) (in [0,1])
  - **d(y)** is an interdependency function that depends on the dynamic functionality and topological attributes of the communication network.
  - **d(y)** should represent the "significance" of the failed communication links on the power grid
  - We represent "importance" by the:
    - Maximum degree of failed nodes
    - Minimum hop-distance between failed nodes and the central node.

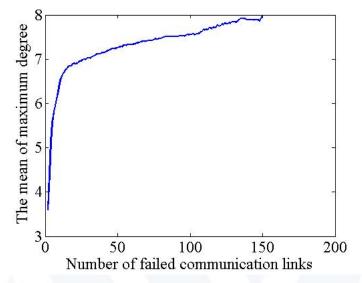
#### Role of communication/control topology

- Optimal power-flow simulations suggest that communication-link failure can be attributed to two main connectivity and topological factors:
  - Minimum hop-distance of the failed communication nodes to the central node
  - Maximum degree of failed communication nodes
- Hence, we can propose:
  - Interdependency variable, d, to be a weighted sum of two probabilities:
    - $p_{hop}^{fail}(h_n)$ : probability of communication-link failure resulting from the state of the connectivity to the central node (hop distance of the failed lines to the central node)
    - $p_{deg}^{fail}(d_n)$ : probability of communication-link failure resulting from the state of the degree of failed communication nodes
    - $d = w \, p_{hop}^{fail}(h_n) \, + \, (1-w) \, p_{deg}^{fail}(d_n)$ ; w is a weight factor between 0 and 1.

### Role of communication/control topology (cont.)

- Optimal power-flow simulations suggest a relationship between:
  - Maximum degree of failed nodes in communication network and number of failed links in communication network.
  - Minimum hop-distance between central control node and failed nodes in communication network.





Parametric approximation

$$p_{hop}^{fail}(h_n) = \begin{cases} \frac{a_1}{h_n^4} + \epsilon \\ \epsilon \end{cases}$$

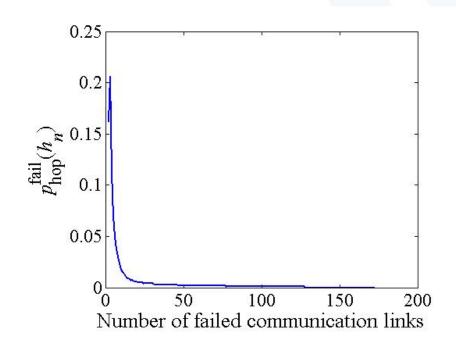
$$1 \le h_n \le m$$
$$h_n > m$$

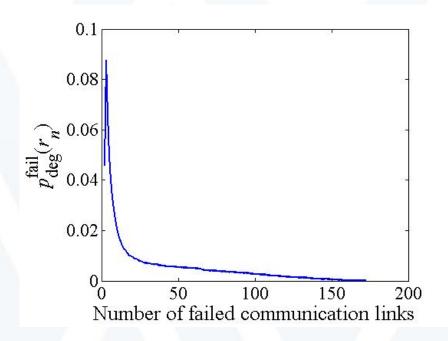
$$1 \le h_n \le m$$

$$h_n > m$$

$$p_{degree}^{fail}(d_n) = \begin{cases} \epsilon & 1 \le d_n < n \\ a_2 d_n^4 + \epsilon & d_n \ge n \end{cases}$$

## Role of communication/control topology (cont.)





Hence, we can represent the interdependency variable

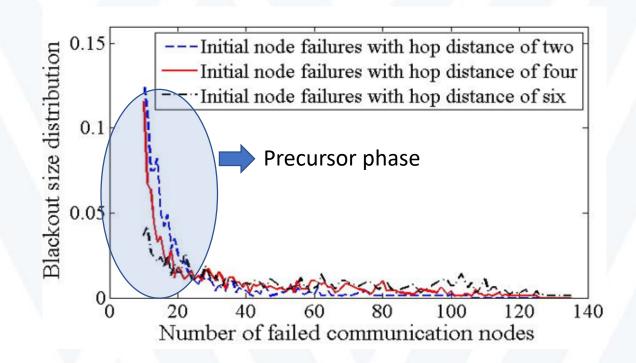
$$d = w p_{hop}^{fail}(h_n) + (1-w) p_{deg}^{fail}(d_n)$$

as

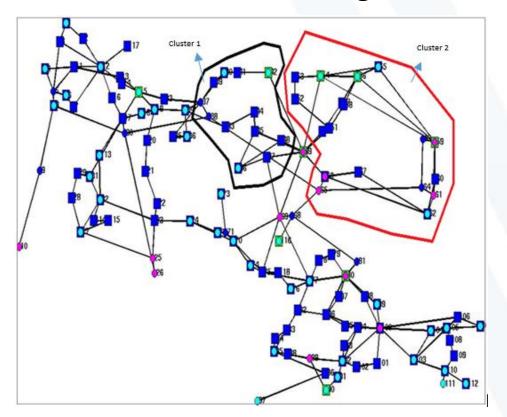
$$d(y_n) = w p_{hop}^{fail}(y_n) + (1-w) p_{deg}^{fail}(y_n)$$

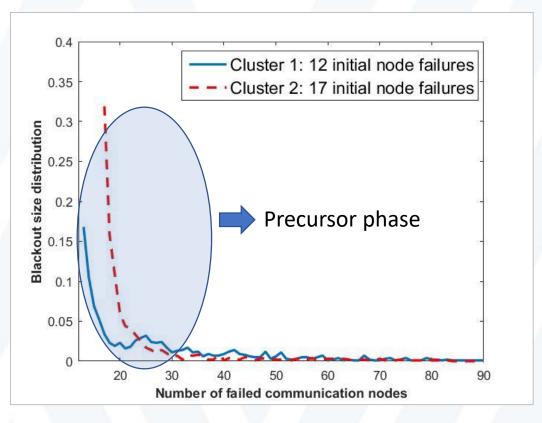
## Blackout distribution comparison for ten initial communication node and power line failures having different minimum hop distances

- We have simulated the blackout distribution in power grid for ten initial communication node and power line failures
- We observed that failure in communication nodes with lower hop distances has higher blackout distribution as failures increase
- When the mean of the minimum hop distance is lower, power grid is more conducive to cascading failures in precursor phase



## Blackout distribution comparison for two clusters having different minimum hop distance & maximum node degree

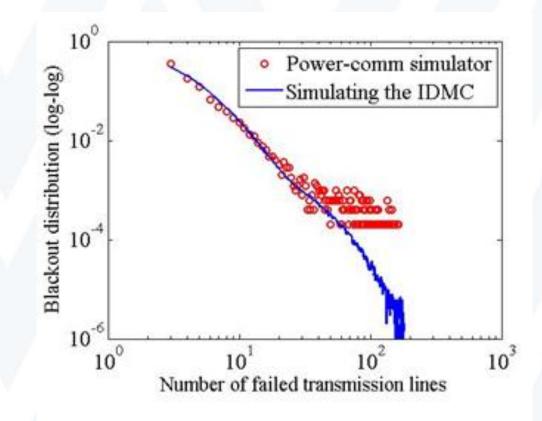




- Both the mean of the minimum hop and the mean of the maximum degree are higher in cluster two to those in cluster one
- Cluster two is more conducive to cascading-failure than cluster one with higher blackout distribution during the precursor phase of the communication network node failures

#### Simulating the Markov chain of the proposed model

- We simulated the Markov chain of the proposed model to validate the IDMC model by comparing its results to those obtained from the coupled communication and power-grid simulator.
- Two results agree in showing a similar trend in the blackout size distribution
- The results obtained from coupled simulator is not precise when the number of failed transmission-lines is large (e.g. over 100), which is due to the limited sample size of large blackouts
- Results validate that the proposed model is effective in capturing the impact of the interdependency between the power system and communication network on cascading-failures in the power grid.

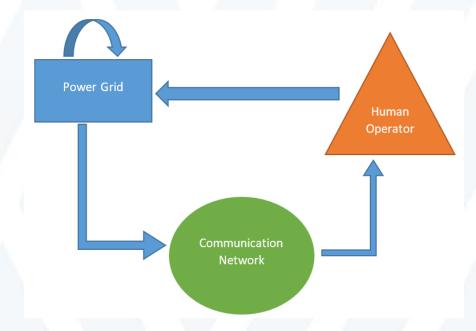


#### **Conclusions**

- We proposed a communication-power interdependency function, d that is determined by hop distance from a central node and the degree of the node in the communication network
- It captures the influence of communication network on the power system under different stress levels
  of the power grid during cascading failures.
- We devised a coupled power-communication simulator and conducted extensive simulations to validate the proposed model
- Blackout probability in the power grid can be significantly impacted by the failures in the communication network when the power grid is under stress.
- The computational time for simulating the proposed model is reduced by a factor of 10e7 to the time
  using the coupled simulator.

## **Ongoing Works**

- Develop a comprehensive 3-layer Markov chain based model to characterize cascading failure in power grid with communication network and human operator error in the loop.( accepted in IGESSC 2017)
- Characterize the impact of initial failures in power grid due to natural disaster, WMD's
- Analyze the impact of lost capacity of the failed transmission lines during cascading failure



#### References

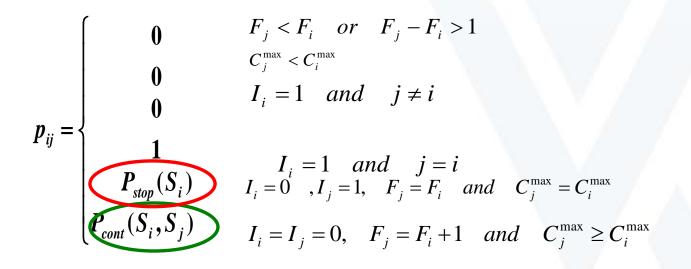
- [1] M. Amin and P. F. Schewe, "Preventing blackouts," Scientific American, vol. 296, no. 5, pp. 60–67, 2007.
- [2] A. Veremyev, A. Sorokin, V. Boginski, and E. L. Pasiliao, "Minimum vertex cover problem for coupled interdependent networks with cascading failures," European Journal of Operational Research, vol. 232, no. 3, pp. 499–511, 2014.
- [3] <a href="https://energy.gov/oe/downloads/blackout-2003-final-report-august-14-2003-blackout-united-states-and-canada-cause-and-canada-causes-and-canada-cause-and-canada-cause-and-canada-cana
- [4] M. Rahnamay-Naeini and M. M. Hayat, "Cascading failures in interdependent infrastructures: An interdependent markov-chain approach," IEEE Transactions on Smart Grid, vol. 7, no. 4, pp. 1997–2006, 2016.
- [5] Rahnamay-Naeini et al., "Stochastic analysis of cascading-failure dynamics in power grids," IEEE Transactions on Power Systems, vol. 29, no. 4, pp. 1767–1779, 2014.

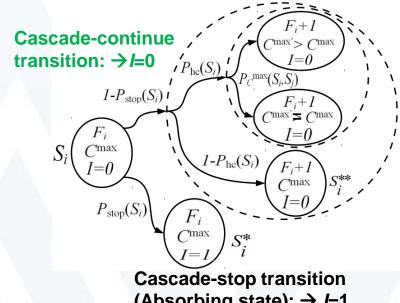
## Thank you for your Attention

**Questions?** 

### **Annex 1: SASE model state transition probabilities**

#### Transition probabilities are state dependent





(Absorbing state):  $\rightarrow$  I=1

### **Annex 2: IDMC model state transition probabilities**

$$f(s_{n+1}|s_n) = \begin{cases} 1 & if \ i_n = 1, x_{n+1} = x_n, l_{n+1} = l_n, \\ y_{n+1} = y_n & \\ q(y_n) & if \ i_n = i_{n+1} = 0, l_n = 0, \\ x_{n+1} = x_n, y_{n+1} = y_n + 1 & \\ 1 - q(y_n) & if \ i_n = i_{n+1} = 0, l_n = 0, \\ x_{n+1} = x_n, y_{n+1} = y_n & \\ 1 - \frac{p(x_n)(1 - d(y_n, h_n, r_n))}{(k_n + (1 - d(y_n, h_n, r_n))(1 - k_n))} & if \ i_n = i_{n+1} = 0, l_n = 1, \\ x_{n+1} = x_n + 1, y_{n+1} = y_n & \\ \frac{p(x_n)(1 - d(y_n, h_n, r_n))}{(k_n + (1 - d(y_n, h_n, r_n))(1 - k_n))} & if \ i_n = 0, i_{n+1} = 1, \\ l_n = 1, x_{n+1} = x_n, y_{n+1} = y_n & \\ 0 & otherwise \end{cases}$$

#### **IDMC** state transitions

Power-communication interdependency: q(x) Communication-power interdependency d(y)

