

Power Blackouts and the Domino Effect:

Real-Life examples and Modeling

Ingve Simonsen

Dep. of Physics, NTNU (Trondheim)

Collaborators: Lubos Buzna, Zilina
 Karsten Peters, Dresden
 Stefan Bornholdt, Bremen
 Dirk Helbing, ETH, Zurich



Acknowledgments

Collaborators:

- Lubos Buzna, Zilina
- Karsten Peters, Dresden
- Stefan Bornholdt, Bremen
- Dirk Helbing, ETH, Zurich
- Rafal Weron, Wroclaw

Financial Support

- EU project **IRRIIS**
 - IRRIIS : “Integrated Risk Reduction of Information-based Infrastructure Systems”
- ERC
 - COST P10, “Physics of Risk”

Motivation

New York, August 14, 2003



Ingve Simonsen

Rome, September 28, 2003

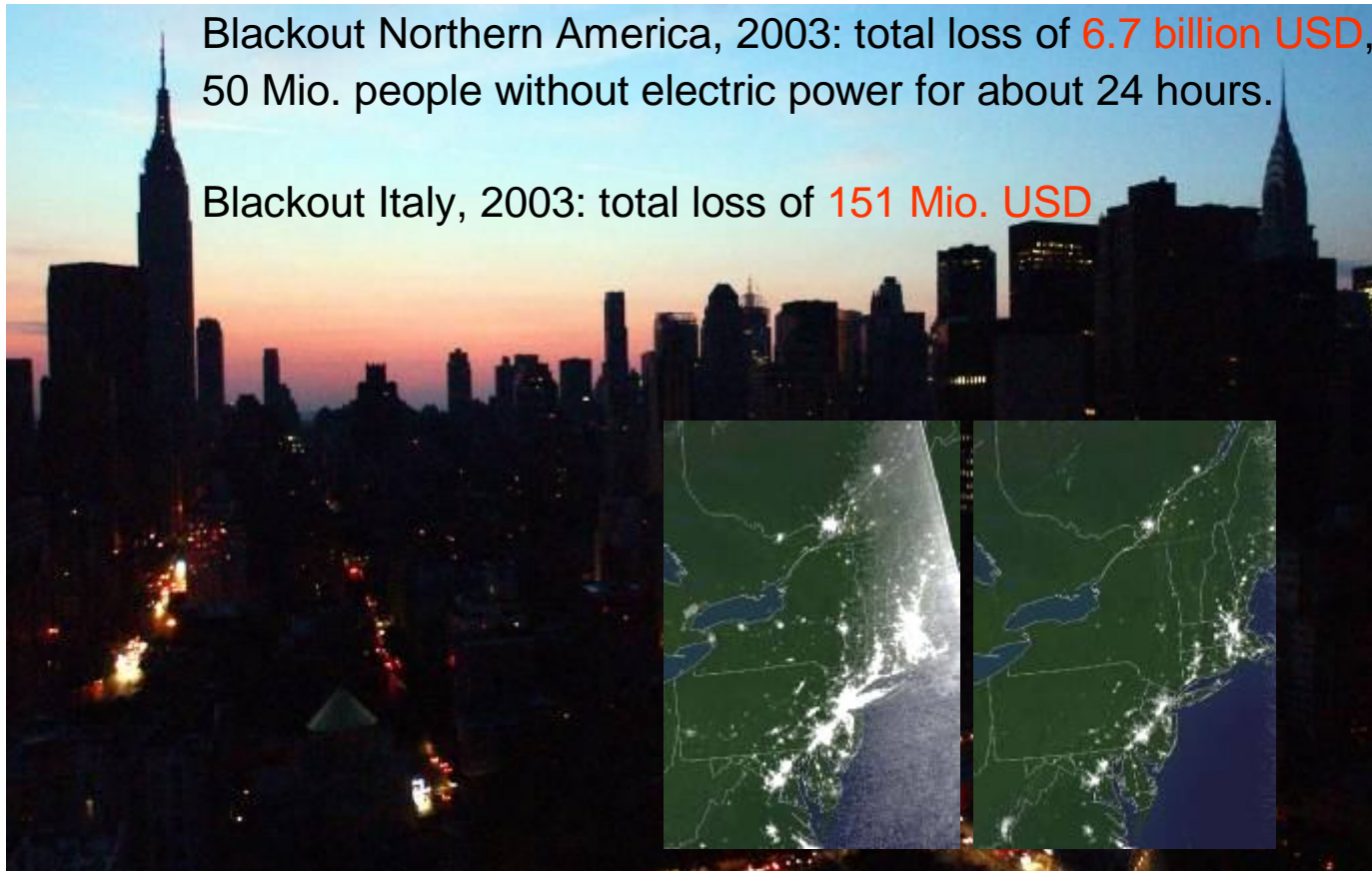


Power Blackouts and the Domino Effect

Motivation

Blackout Northern America, 2003: total loss of **6.7 billion USD**, 50 Mio. people without electric power for about 24 hours.

Blackout Italy, 2003: total loss of **151 Mio. USD**

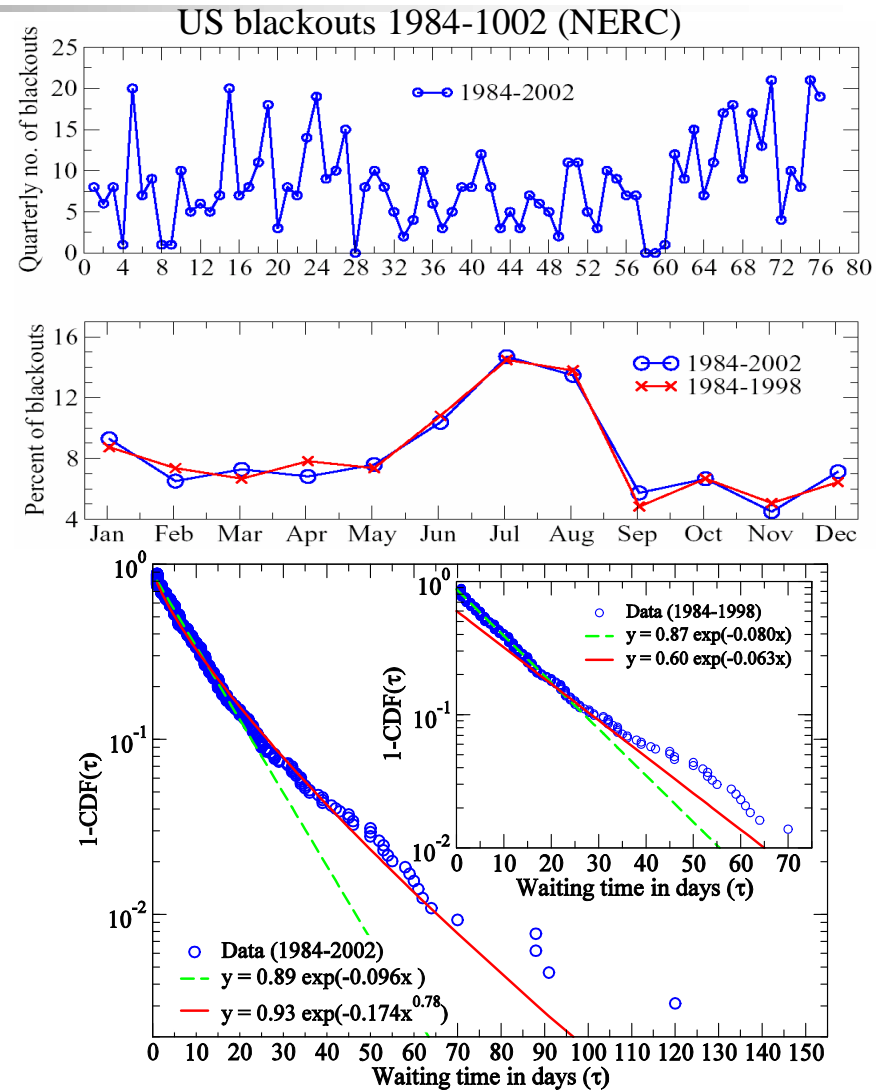


Blackout in parts of the USA and Canada (2003), an impressive example of the long-reaching accompaniments of supply network failures.

Power Blackouts : How frequent are they?

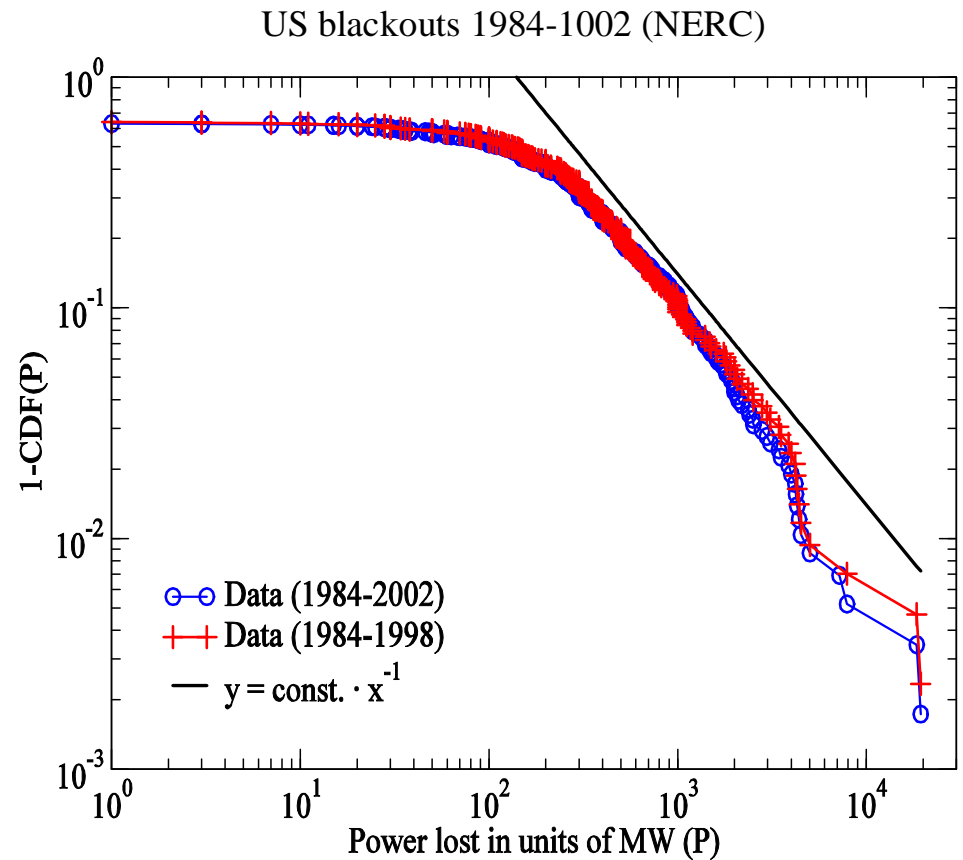
- North American Electricity Reliability Council (NERC) data
 - Analyzed by Carreras, Dobson, Newman & Poole
 - 15 years of data (1984-1998)
 - 427 blackouts
 - on average 28.5 per year, waiting time of 12 days

- Three measures of blackout size
 - energy unserved (MWh)
 - amount of power lost (MW)
 - number of customers affected



Risk of Power Blackouts

- There are rather few *large* blackouts
 - So why should we care at all?
- Risk = Probability * Cost
- Large Power Blackouts are the most RISKY!



Source : Weron and Simonsen (2005)

Power Blackouts: Real-Life examples

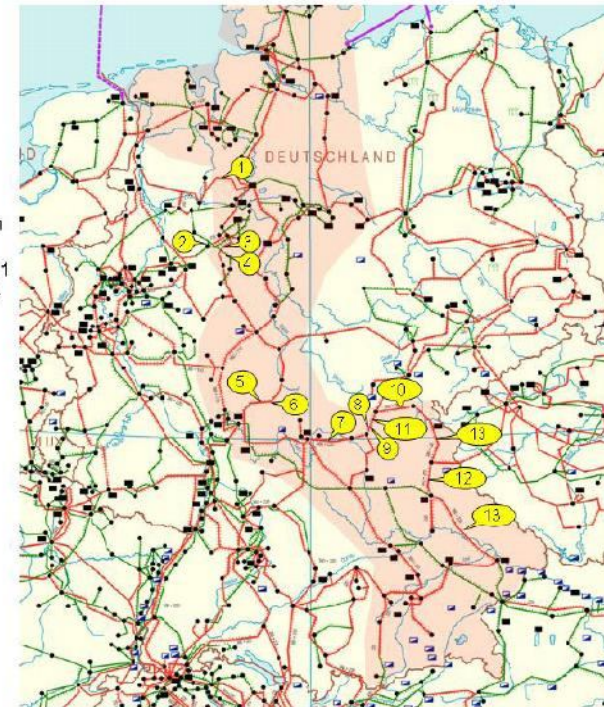
Europe Nov. 2006: What happened...?

State of the power grid shortly before the incident



Sequence of events on November 4, 2006

Nr.	Zeit	kV	Leitung
1	22:10:13	380	Wehrendorf-Landesbergen
2	22:10:15	220	Bielefeld/Ost-Spexard
3	22:10:19	380	Bechterdissen-Elsen
4	22:10:22	220	Paderborn/Süd-Bechterdissen/Gütersloh
5	22:10:22	380	Dipperz-Großkrotzenburg 1
6	22:10:25	380	Großkrotzenburg-Dipperz 2
7	22:10:27	380	Oberhaid-Grafenrheinfeld
8	22:10:27	380	Redwitz-Raitersaich
9	22:10:27	380	Redwitz-Oberhaid
10	22:10:27	380	Redwitz-Etzenricht
11	22:10:27	220	Würgau-Redwitz
12	22:10:27	380	Etzenricht-Schwandorf
13	22:10:27	220	Mechlenreuth-Schwandorf
14	22:10:27	380	Schwandorf-Pleinting



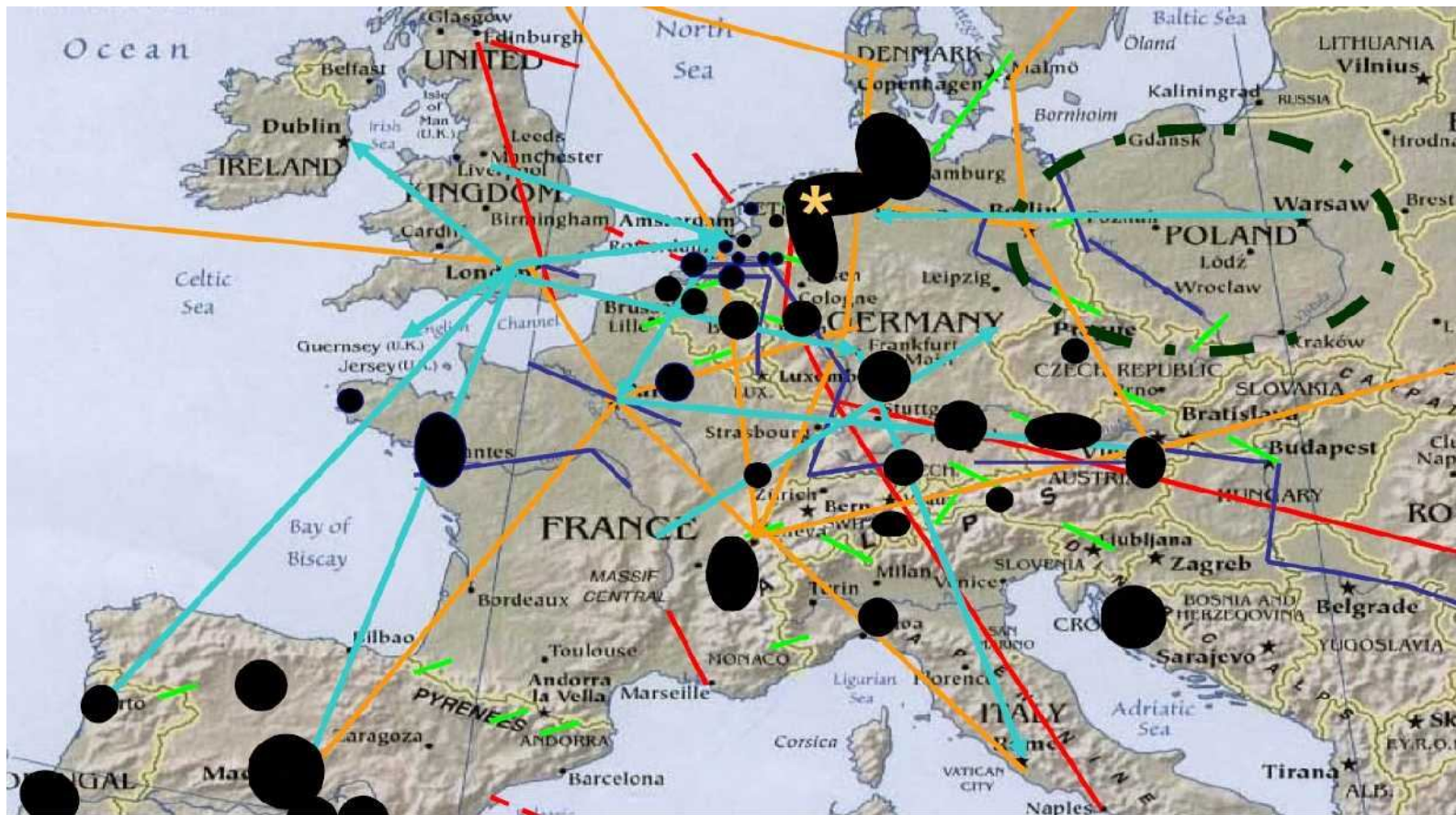
1,3,4,5 – lines switched off for construction work

2 – line switched off for the transfer of a ship by Meyer-Werft

Source : Report on the system incident of November 4, 2006, E.ON Netz GmbH

Power Blackouts: Real-Life examples

Failure in the continental European electricity grid on November 4, 2006



EU project IRRIS: E. Liuf (2007) Critical Infrastructure protection, R&D view

Power Blackouts: The Domino Effect (Cascading Failure)

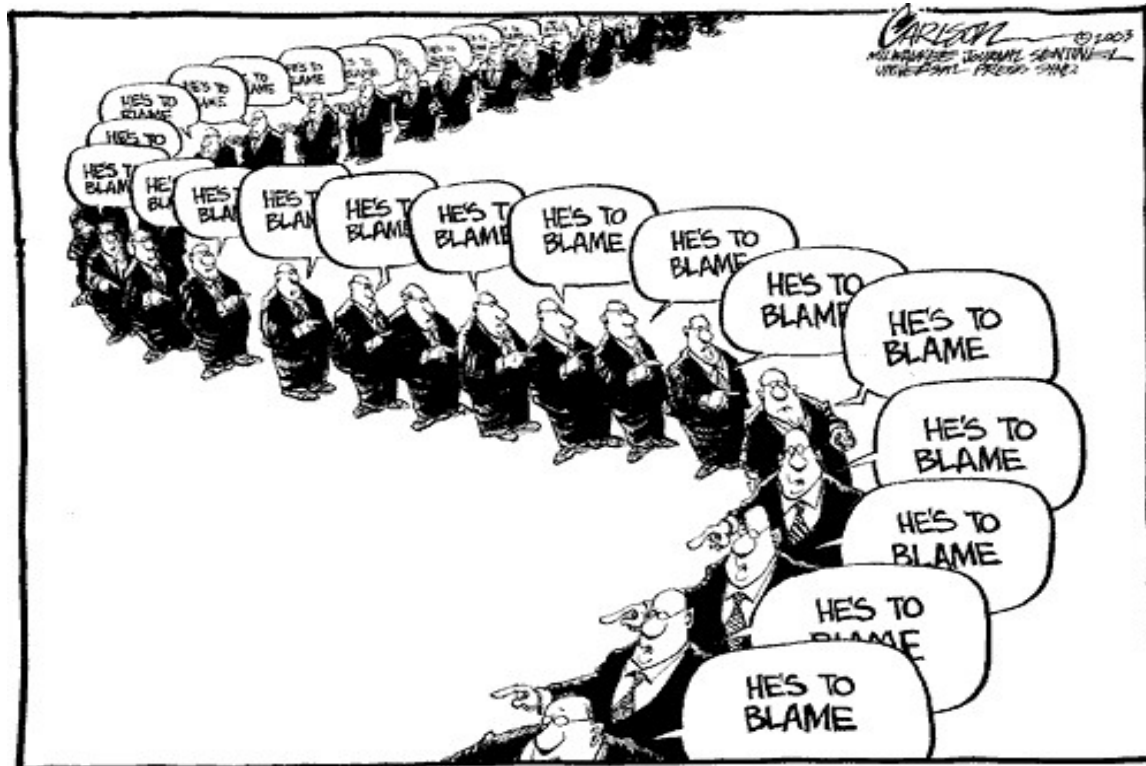


“Under certain conditions, a network component shutting down can cause current fluctuations in neighboring segments of the network, though this is unlikely, leading to a cascading failure of a larger section of the network. This may range from a building, to a block, to an entire city, to the entire electrical grid.”

Source :Wikipedia

Power Blackouts: Real-Life examples

US Blackout Nov. 2003: What happened...?



BLACKOUT OF '03 —The "CASCADE EFFECT"

10/19

See : [Wikipedia](#) for sequence of events



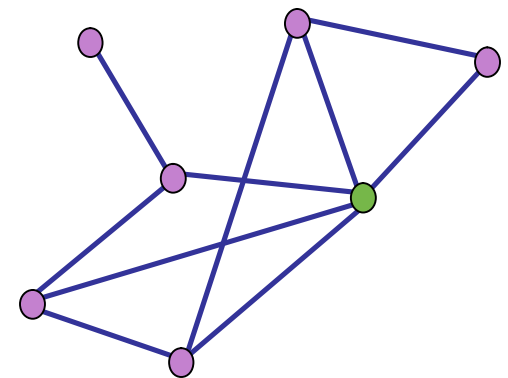
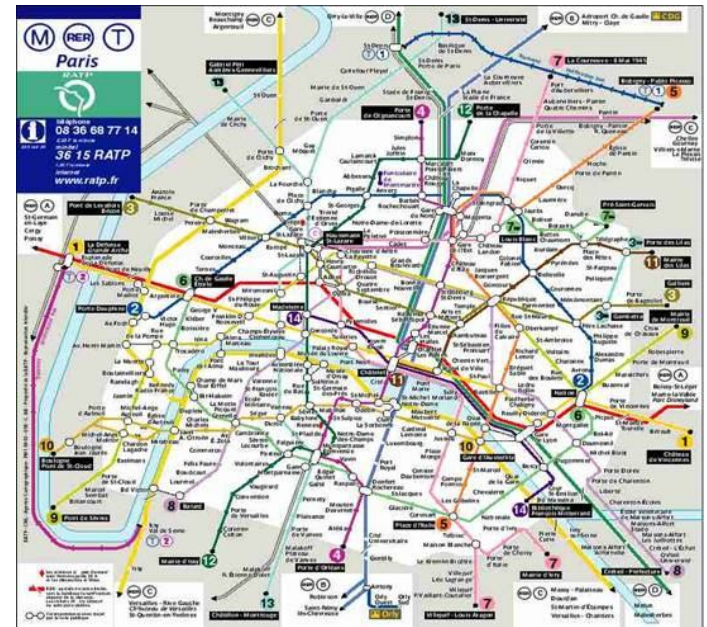
Power Blackouts : Summary

- *Cascading failures* do exist in real life system
 - Examples
 - The power grid
 - Telecommunication networks
 - Transportation systems
 - Computer networks/ the Internet
 - Pipe line systems (water/gas/oil)
- They can be very *costly*
- They typically affect many people

Question : *How* can one protect (supply) network systems against cascading failures?

A Short Primer on Complex Networks

- A network is a collection of
 - Nodes connected by links
- Adjacency matrix W_{ij}
- Degree (#links) distribution
 - Scale-free (e.g. the Internet)
 - Exponential (e.g. power networks)
- Betweenness centrality of a node
 - Total # shortest paths passing through *that* node for any pair of nodes





A few words on System Design

- The systems are designed with a *given load* in mind
- To ensure stability, the engineering approach, is to introduce some *overcapacity* into the system (security margins)
- ...but overcapacity is *costly*!
- System robustness is often ONLY evaluated locally

- Cascading failure: When an initial perturbation occurs, loads have to redistribute. If the resulting loads exceed the capacities of link/nodes, new failures can result.... “the Domino effect”

Why do we have blackouts.....?

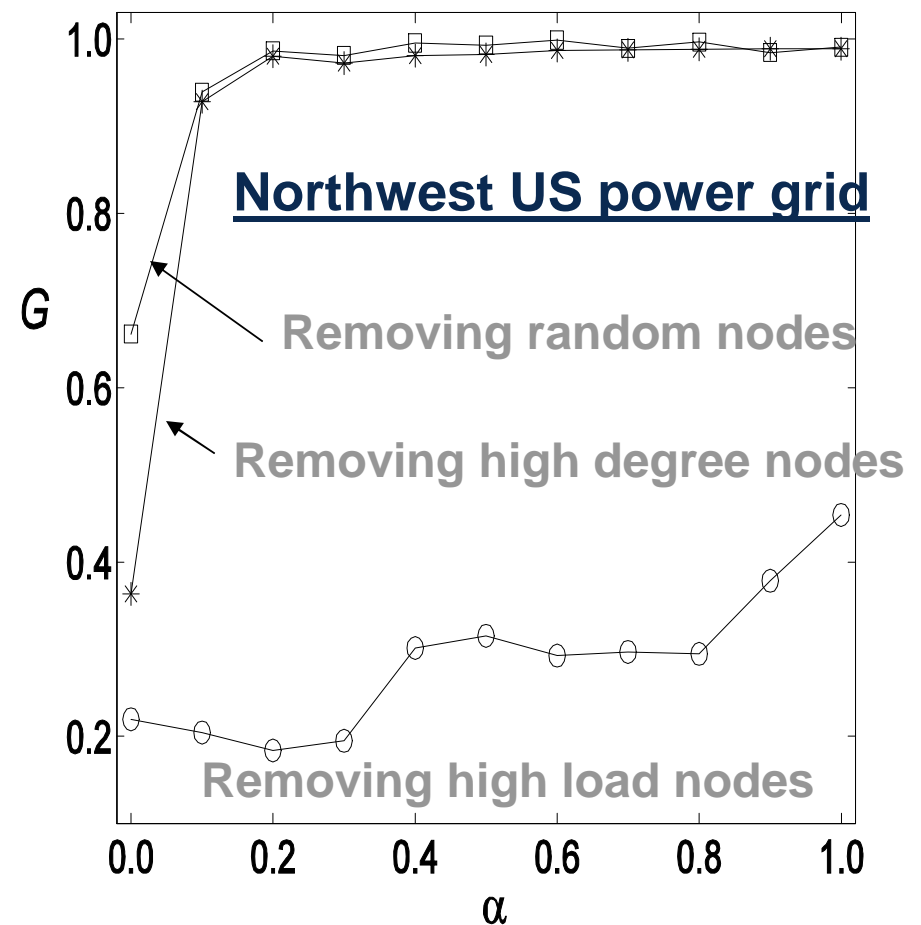
- System load (throughput)
 - optimized to get the maximum out of the system
 - high load means small operating margins
 - has impact on interactions and component failures
- Tradeoff between load and risk of failure
 - at system level
 - for system components
- What is the role of the deregulation?



Previous physics works : Cascading Failures

Seminal paper by Motter and Lai: PRE 66, 065102R (2002)

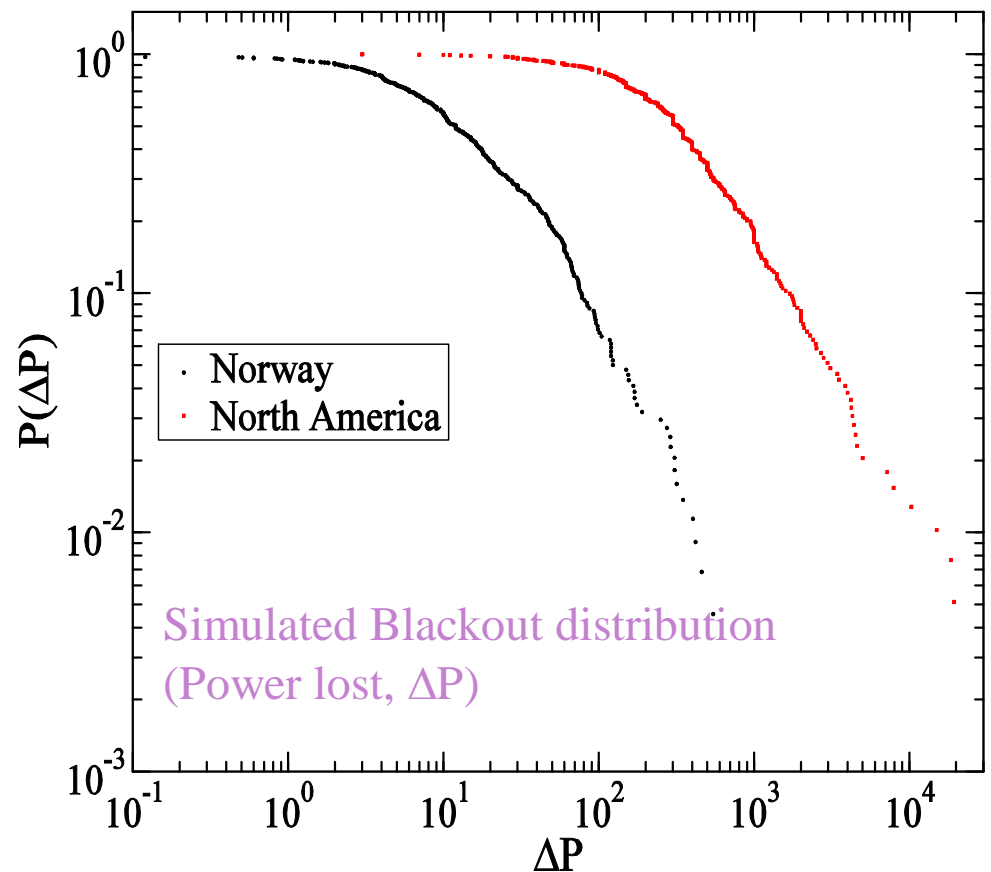
- No sinks/sources
- Initial load of a node, L_i , is its betweenness centrality
- Node Capacity : $C_i = (1+\alpha)L_i$
- One probes only the *stationary state* of the system
- The system is *perturbed*, and the fraction of nodes remaining in the largest component, G , is recorded after the cascade has stopped.



Previous works : Cascading Failures

Bakke *et al.* Europhys. Lett **76**, 717 (2006)

- More physically realistic model for the current flow (the Kirchoff laws)
- “*The price to pay*” : one has to solve a large system of linear eq.
- NOTE: Also here one probes only the *stationary state* of the system



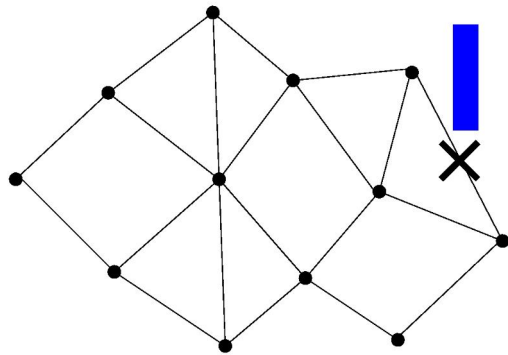


Previous works : Summary/Open Questions

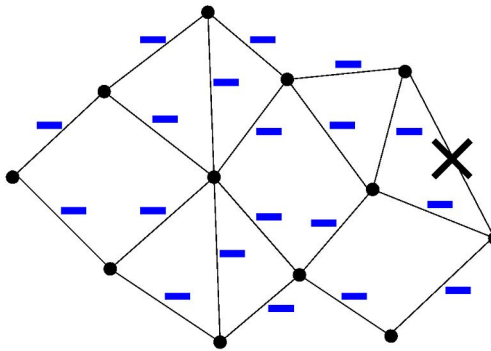
- Previous works of cascading failures exclusively considered the stationary state
- We asked ourselves: Why should the system *not* experience additional failure due to overloading during the transient period?
- Question to address:
 - What is the role played by dynamics in cascading failures in complex networks
- A dynamical model is needed for such a study
 - But which one to choose?

Expected difference between a static and a dynamic model for flow redistribution

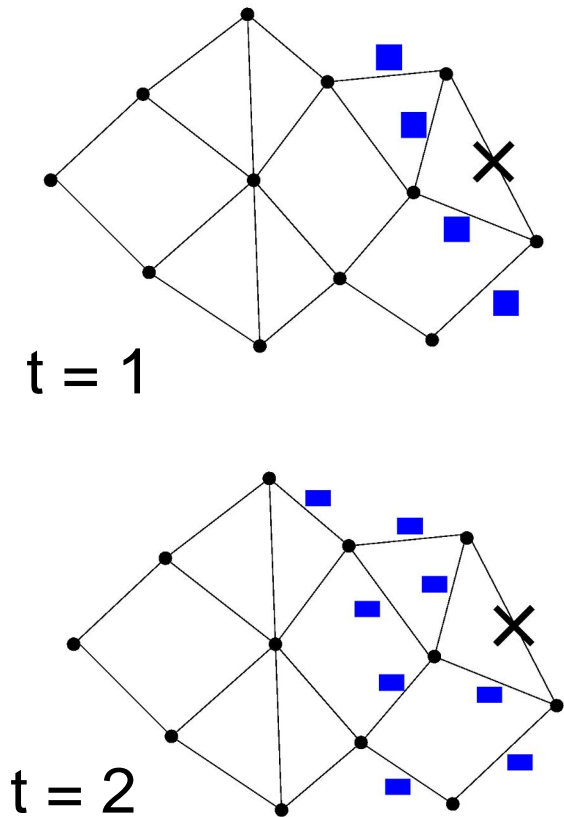
Initial failure



Stationary model



Dynamic model



Model : Requirements

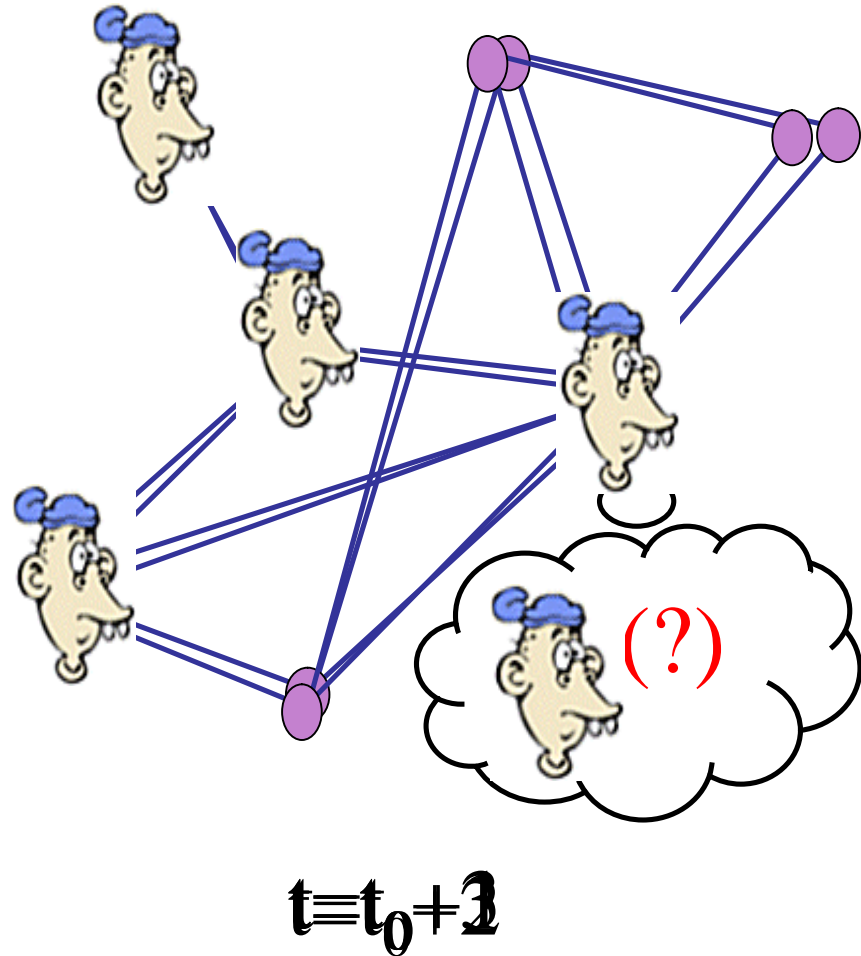
- It should be:
 - **Generic** : no particular physical process is addressed
 - As simple as possible, but not simpler...
- Important ingredient (in our opinion)
 - The flowing quantity should be CONSERVED



Our solution : A Random Walk (or Diffusion type) model !

The Dynamical Model: Basic Principle (Flow/Diffusion Model)

- Random walkers (i.e. particles) “live” on the nodes
- They are moving (flowing) around on the network!
- In each time step, a walker move *one* step forward towards one of the neighboring nodes chosen by random
- This process is repeated over and over again.....
- **Note:** The number of walkers is constant in time



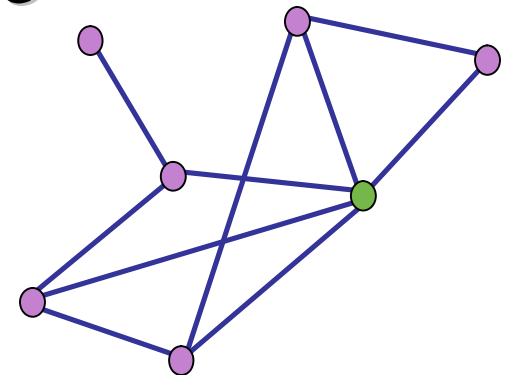
The Dynamical Model: The Master Equation

- Convention : W_{ij} refers to the link from node j to i ;
- Define the outgoing link weight from node j : $w_j = \sum_i W_{ij}$
- The change in no. of particle at node i from t to $t+1$

$$n_i(t+1) - n_i(t) = \sum_j W_{ij} \frac{n_j(t)}{w_j} - \sum_j W_{ji} \frac{n_i(t)}{w_i} + n_i^\pm(t),$$

- The “outgoing-term” is simple, and one gets

$$n_i(t+1) = \sum_j W_{ij} \frac{n_j(t)}{w_j} + n_i^\pm(t)$$





The Dynamical Model: The Master Equation

- Define the relative fraction of walkers (total N) at node i :

$$\rho_i(t) = \frac{n_i(t)}{N}$$

- The outgoing current per weight unit from node i is:

$$c_i(t) = \frac{\rho_i(t)}{w_i} = \frac{n_i(t)}{w_i N}$$

- Hence, it follows

$$c(t+1) = Tc(t) + j^\pm(t); \quad T_{ij} = \frac{W_{ij}}{w_i}$$



The Dynamical Model : Summary

- Our simple dynamical model incorporates:
 - Flow conservation
 - Network topology
 - Load redistribution

$$c(t+1) = Tc(t) + j^\pm(t); \quad T_{ij} = \frac{W_{ij}}{w_i}$$

$c_i(t)$: The outgoing current from node i per link weight unit

The Dynamic Model

Network:

\mathcal{N} set of nodes

\mathcal{L} set of links

W adjacency matrix ($W_{ij} \geq 0$, link weight)

Model dynamics:

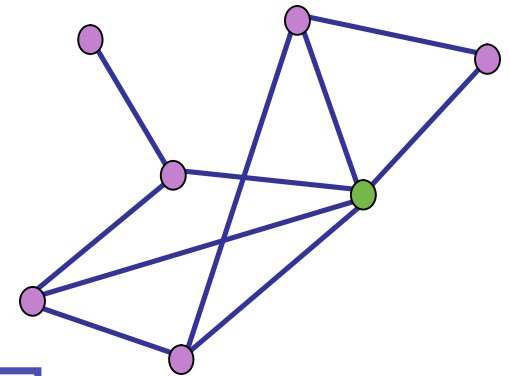
$$n_i(t + 1) = \sum_{j=1}^{\mathcal{N}} T_{ij} n_j(t) + n_i^{\pm}$$

(Master equation)

$n_i(t)$ number of particles hosted by node i at the time t

$$T_{ij} = W_{ij}/w_j, \quad w_j = \sum_{\ell=1}^{\mathcal{N}} W_{\ell j}$$

$n_i^{\pm} > 0$ node is source, $n_i^{\pm} < 0$ node is sink



Ref : I. Simonsen, L. Buzna, K. Peters, S. Bornholdt, D. Helbing, Phys. Rev. Lett. **100**, 218701 (2008)

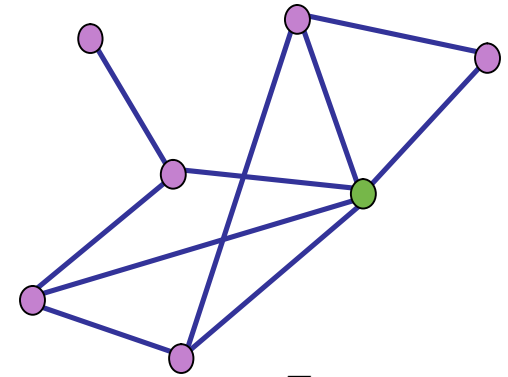
Stationary and Dynamic Models of Cascading Failures

Model normalization:

$$\rho_i(t) = n_i(t)/N \quad \text{nodal particle density}$$

$$c_i(t) = \rho_i(t)/w_i \quad \text{utilization of outflow current}$$

$$j_i^\pm = n_i^\pm / (Nw_i) \quad \text{sinks and sources terms}$$



Dynamic model

$$\mathbf{c}(t+1) = \mathcal{T}\mathbf{c}(t) + \mathbf{j}^\pm, \quad (\tau = T^T)$$

$$c_i^{(0)}(\infty) = 1/(Nw_i) \quad \text{stationary solution for } j^\pm = 0, \text{ otherwise}$$

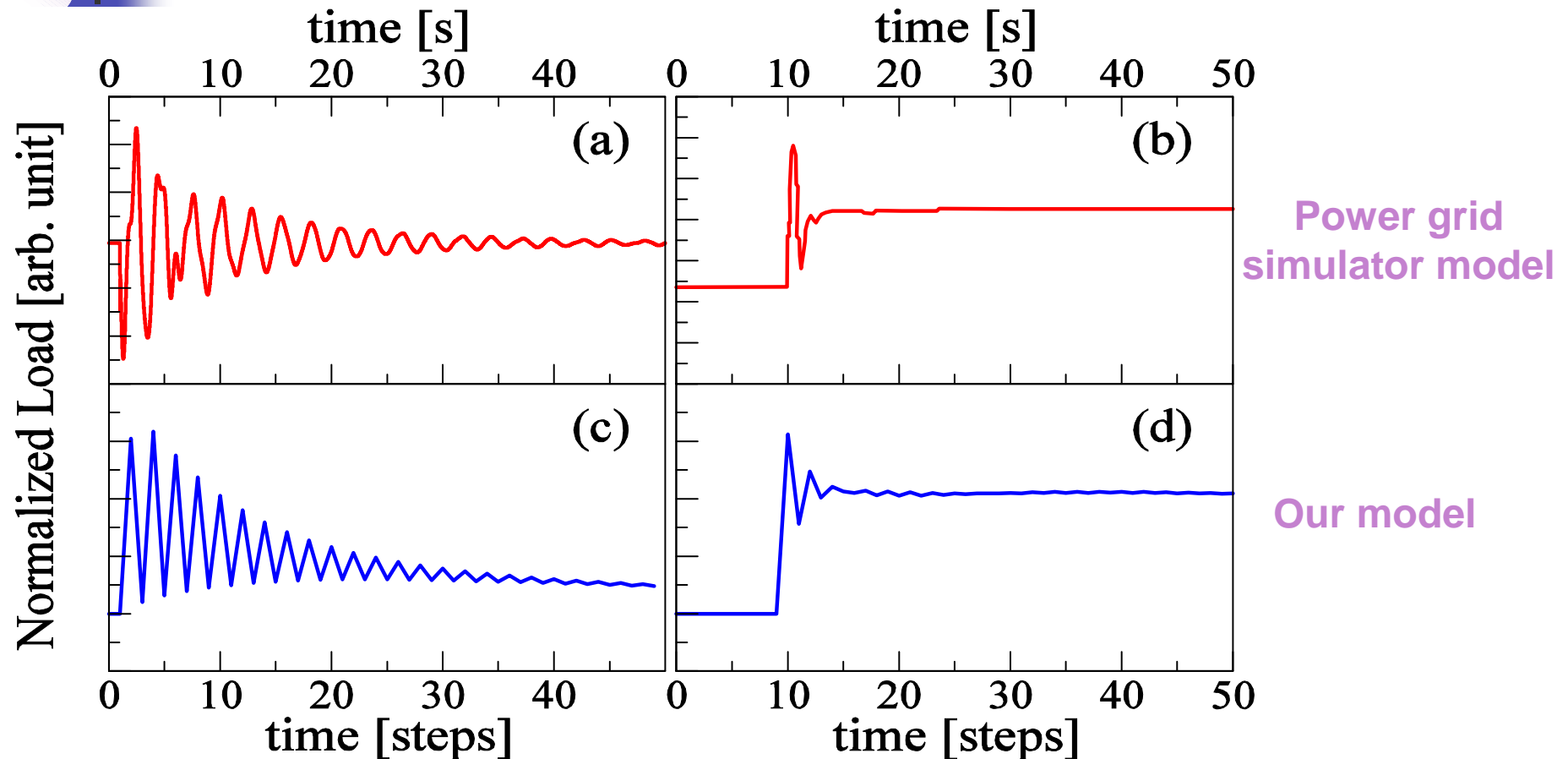
Stationary model

$$\mathbf{c}(\infty) = \mathbf{c}^{(0)}(\infty) + (\mathbf{1} - \mathcal{T})^+ \mathbf{j}^\pm$$

$(\mathbf{1} - \mathcal{T})^+$ generalized inverse of matrix $\mathbf{1} - \mathcal{T}$ Link flow:

$$C_{ij}(t) = W_{ij}c_j(t) \quad \text{current on link from } j \text{ to } i \quad L_{ij}(t) = C_{ij}(t) + C_{ji}(t)$$

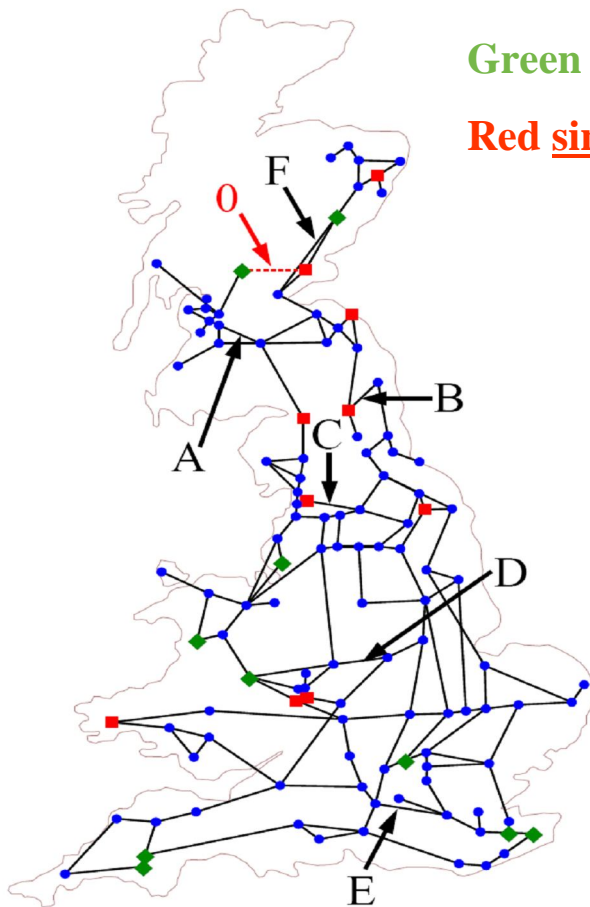
Model Dynamics: Is it realistic?



Source : R. Sadikovic, Power flow control with UPFC, (internal report)

EUROSTAG power simulator : www.aurostag.epfl.ch

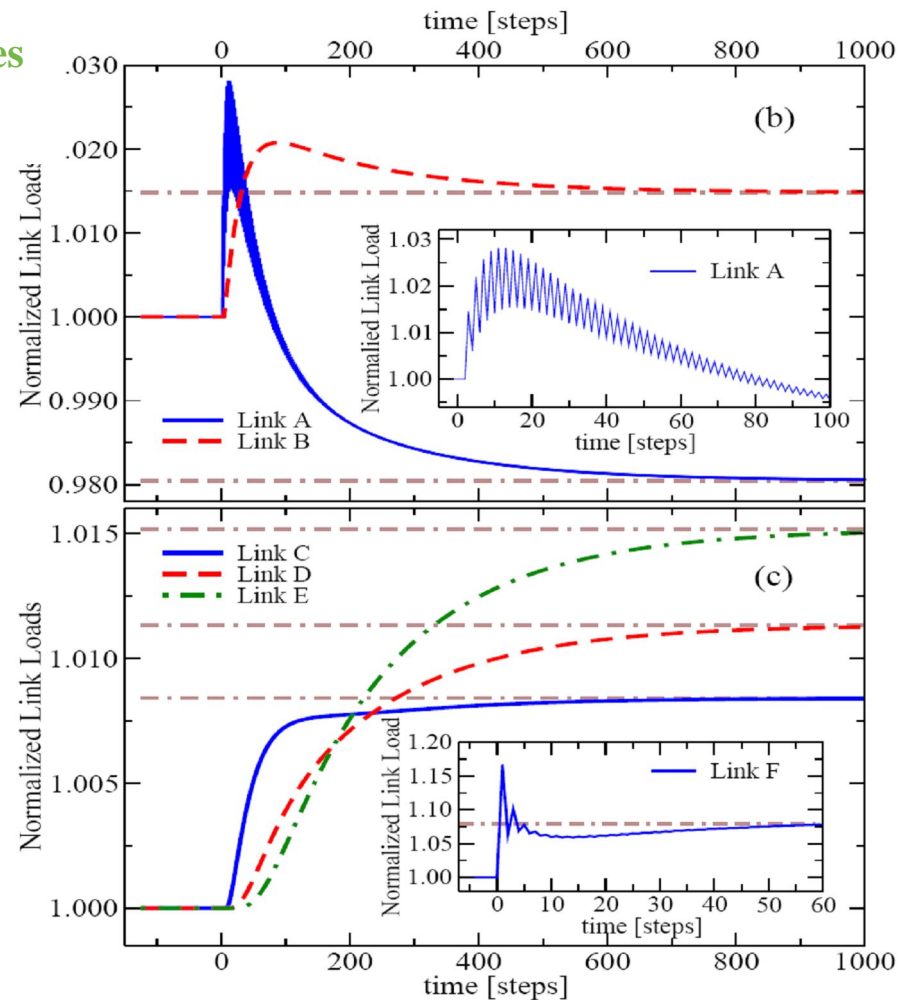
Model Dynamics: UK high voltage power grid (300-400kV)



Green source nodes

Red sink nodes

At $t=0$, link 0 is broken!



When does a link/node fail?

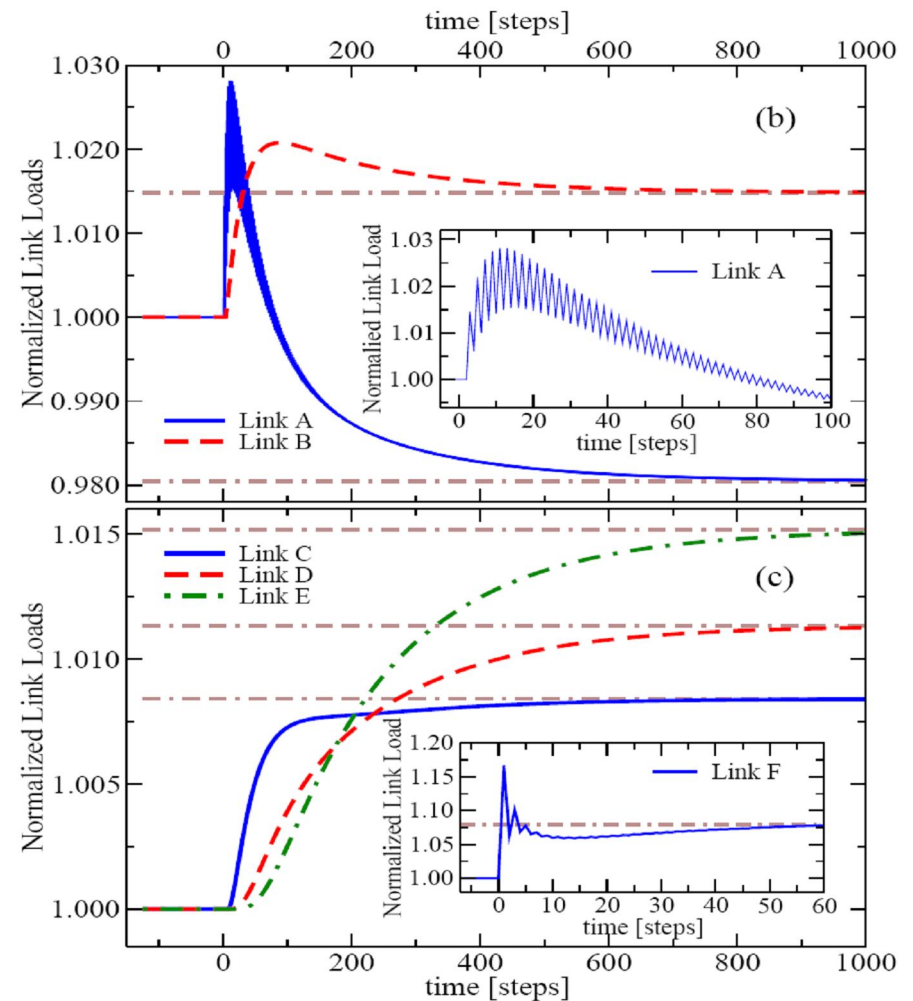
- Link/node capacities relative to the undisturbed state (L_{ij}) via a *tolerance parameter* α

$$C_{ij} = (1 + \alpha) L_{ij},$$

- A link/node fails whenever its *current* load, $C_{ij}(t)$ exceeds the capacity of *that* link/node

Failure if :


$$C_{ij}(t) > (1 + \alpha) L_{ij}$$





Main steps of the simulations

The simulations consist of the following steps:

- 
1. A *triggering event* ($t=0$) [remove a random link]
 2. Calculate the [new] link loads $C_{ij}(t)$
 3. Check if any links are *overloaded* via $C_{ij}(t) > (1 + \alpha)L_{ij}$
 1. If so *remove* such overloaded links
 4. Repeat step 2 and 3 till no more links are overloaded
 5. Average the results over the triggering event of pnt. 1 (and source and sinks locations)

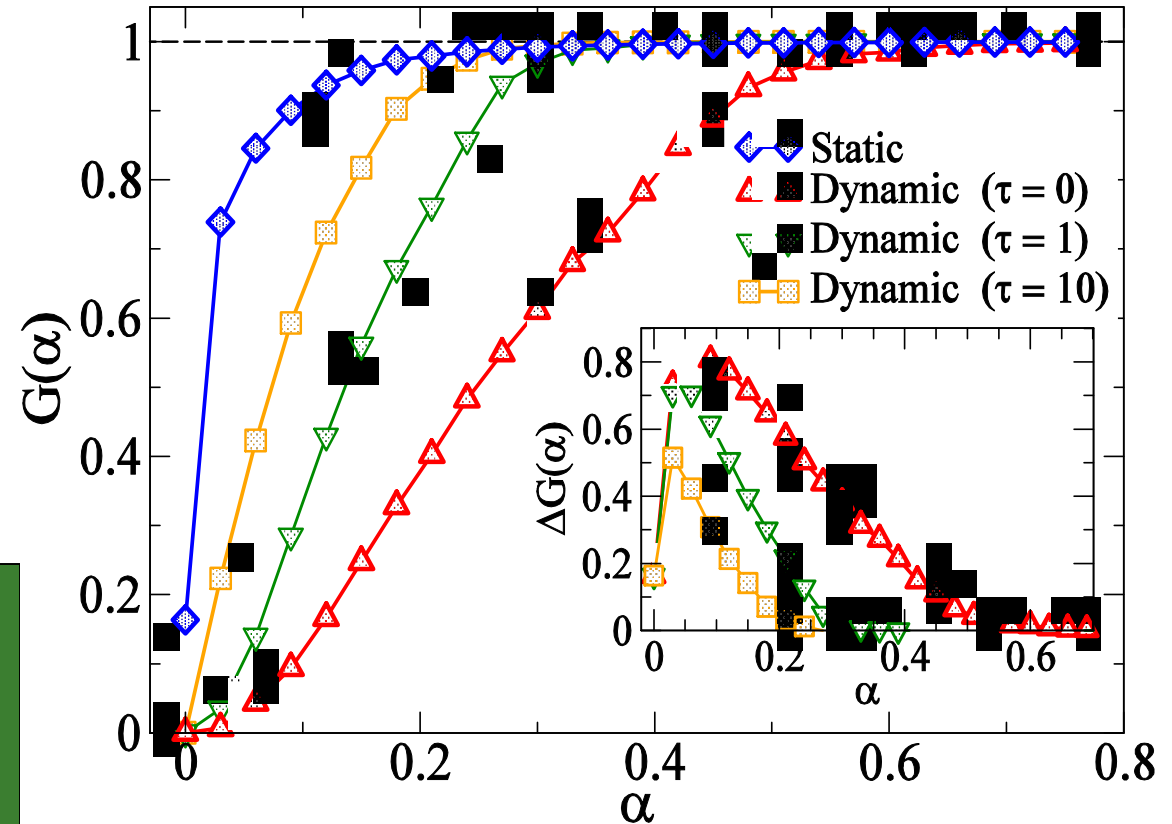
Stationary Model vs. Dynamic Model :

The northwestern US power transmission grid

$|\mathcal{N}|$ number of nodes (5000)
 $|\mathcal{L}|$ number of links
 $|\mathcal{N}_R|$ number of remaining nodes
 $|\mathcal{L}_R|$ number of remaining links

The stationary model can *overestimates robustness* by more than 80% (in this case)

Overload exposure times may be relevant and will increase the robustness.....



$$G_{\mathcal{L}}(\alpha) = \frac{|\mathcal{L}_R|}{|\mathcal{L}|} \approx G_{\mathcal{N}}(\alpha) = \frac{|\mathcal{N}_R|}{|\mathcal{N}|} = G(\alpha)$$



Stationary Model vs. Dynamic Model :

The role of the two time-scales

- There are two characteristic time-scales in the problem:
 - Overload exposure time (protection system response time): τ
 - Typical transient time for the dynamics: τ_0
- Control parameter : $\chi = \frac{\tau}{\tau_0}$
 - Static cascading failure model: $\chi \gg 1$
 - Dynamical ($\tau=0$) cascading failure model : $\chi=0$
- The real situation is probably somewhere in between....

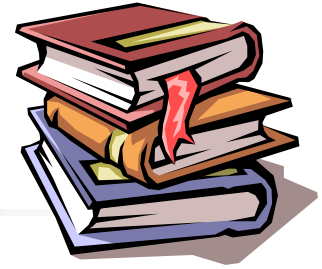
Conclusions



- *The dynamical process on the network is important to consider when evaluating network robustness (cascading)*
 - Using a stationary model may dramatically overestimate (by 80-95%) the robustness of the underlying network
 - The actual overestimation do depend on the actual overload exposure time
- In a dynamical model:
 - links may fail that otherwise would not have done so (overshooting)
 - The proximity to a disturbance is more important in a dynamical model

Thank you for your attention!

References



- Dynamical model :

- I. Simonsen, L. Buzna, K. Peters, S. Bornholdt, D. Helbing, Phys. Rev. Lett. **100**, 218701 (2008).
- See also :
 - Phys. Rev. Lett. **90**, 14870 (2003).
 - Physica A **357**, 317 (2005).
 - Physica A **336**, 163 (2004).

- Stationary models:

- Motter and Lai, Phys. Rev. E **66**, 065102R (2002).
- Bakke *et al.* Europhys. Lett. **76**, 717 (2006).