

Assessing the robustness of the U.S. power grid under extreme wind events

Tomas Scagliarini Mauro Faccin and Manlio De Domenico
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CCS Italy 2025

Motivation

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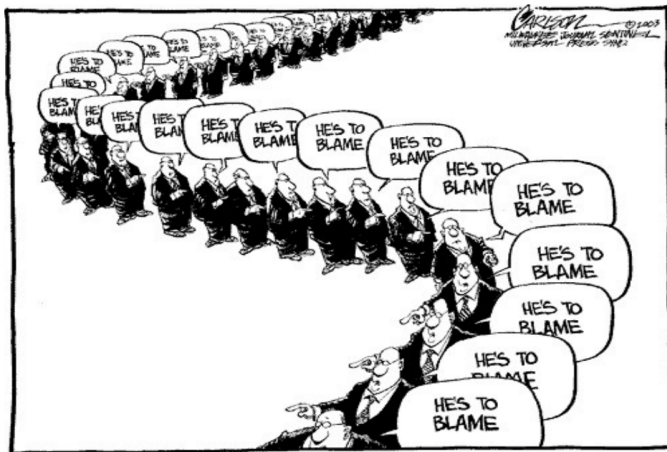


Iberian Peninsula blackout (2025)



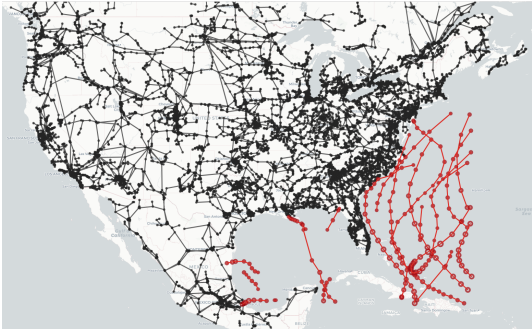
Italy blackout (2003)

What happened?



BLACKOUT OF '03 —The "CASCADE EFFECT"

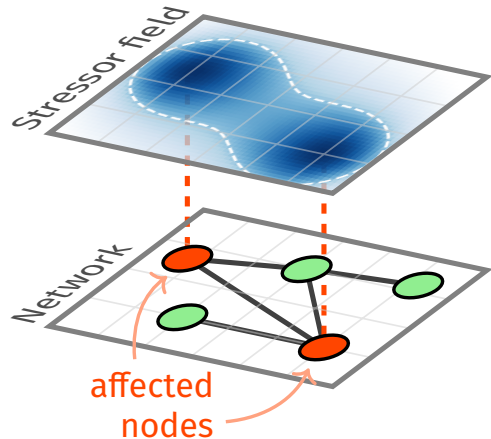
Weather is a primary cause of disruptions



Hurricane paths and US grid

Modelling

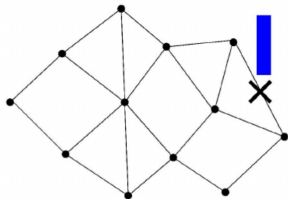
- Can we build a **dynamical model** of cascade spreading?
- How to model the impact of **exogenous events** (e.g. strong winds)?
- Can we validate using real weather data and historical blackouts?



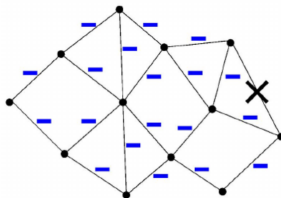
The basic framework

Flow redistribution in a power grid

Initial failure

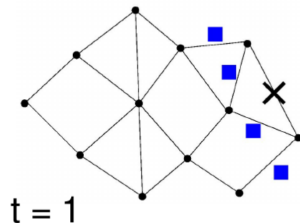


Stationary model

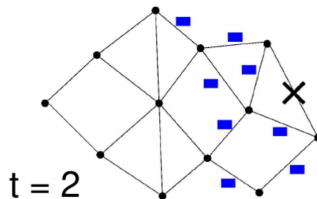


e.g. Motter model

Dynamic model



$t = 1$



$t = 2$

Flow conserving model

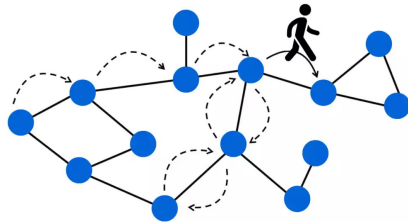
We can imagine the current flow with a **random-walk** like model.

$$n_j(t+1) - n_j(t) = \underbrace{\sum_i \frac{W_{ij}}{k_i} n_i(t)}_{\text{inflow}} - \underbrace{\sum_i \frac{W_{ji}}{k_j} n_j(t)}_{\text{outflow}} + n_j^\pm(t)$$

where $k_i = \sum_j W_{ij}$ is the **degree** and $T_{ij} = \frac{W_{ij}}{k_i}$ the **transfer matrix** from i to j .

Basic ingredients:

- ✓ Flow must be conserved!
- ✓ Flow redistribution
- ✓ Initial failures triggered by weather



Flow conserving model

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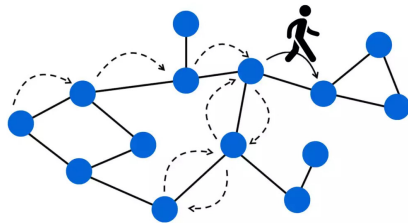
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De Groot model

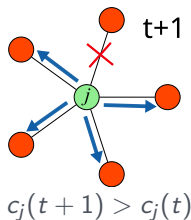
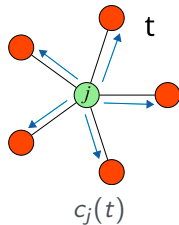
By introducing the **density** of walkers $\rho_j(t) \equiv \frac{n_j(t)}{M}$

$$\rho_j(t+1) = \sum_i \frac{W_{ij}}{k_i} \rho_i(t) + \rho_j^\pm(t)$$

We introduce $c_j(t) \equiv \frac{\rho_j(t)}{k_j}$ the **outgoing current** per unit weight

$$c_j(t+1) = \sum_{i=1}^N \frac{W_{ij}}{k_j} c_i(t) = \sum_i Q_{ji} c_i(t)$$

Note that the transfer matrix $Q_{ji} = \frac{W_{ij}}{k_j} = (T_{ij})^\top$ is the **transpose** of a random walk.



Stationary solution

This model was introduced to describe **opinion dynamics** (De Groot model, 1974)
In vector form, and inserting **source/sink** terms $\vec{j}^{\pm}$

$$\vec{c}(t+1) = \vec{c}(t)\hat{Q} + \vec{j}^{\pm}$$

Stationary solution

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In vector form, and inserting **source/sink** terms $\vec{j}^{\pm}$

$$\vec{c}(t+1) = \vec{c}(t)\hat{Q} + \vec{j}^{\pm}$$

If $\vec{j}^{\pm} = 0$ then the stationary solution is a constant vector $c_i^{(0)}(\infty) \sim \frac{1}{\sqrt{N}}$

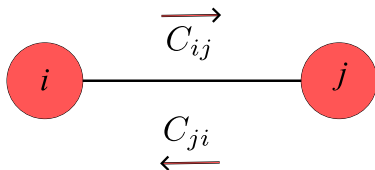
If $\vec{j}^{\pm} \neq 0$, it can be expressed as

$$\vec{c}(\infty) = \vec{c}^{(0)}(\infty) + (\vec{1} - \vec{Q})^+ \vec{j}^{\pm}$$

When does a link fail?

Total **directed current** on link $i \rightarrow j$ becomes

$$C_{ij}(t) = W_{ij}c_j(t)$$



from which we can compute the **total current** on link $j \leftrightarrow i$

$$L_{ij}(t) = C_{ij}(t) + C_{ji}(t)$$

When does a link fail?

Maximum capacity $\mathcal{M}$ is related to the initial load (L_{ij}^0) via a tolerance parameter $\alpha > 0$

$$\mathcal{M}_{ij} = (1 + \alpha)L_{ij}^0$$

A link fails whenever its current load L_{ij} exceed the capacity of that link

Link $i - j$ fails if

$$L_{ij}(t) > \mathcal{M}_{ij}$$

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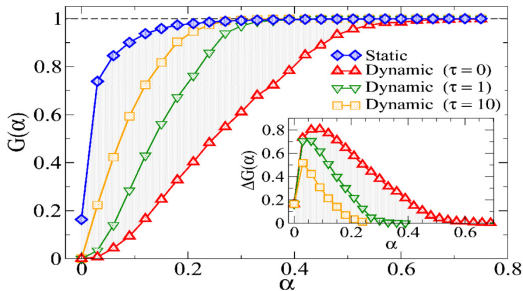
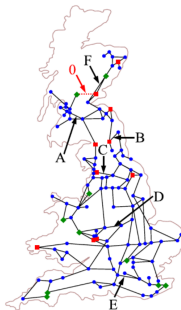
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A second discrete parameter is the overload exposure time τ
 $\tau > 0$: the system will have to be overloaded for a certain time before causing a failure.

Simulation on the UK power grid



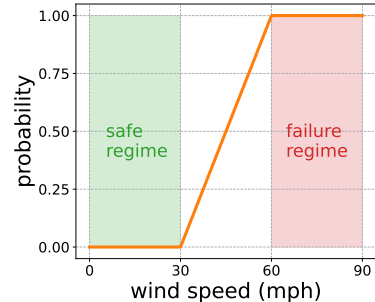
$$G(\alpha) = \frac{\mathcal{N}(\text{Survived links})}{\mathcal{N}(\text{Links})}$$

(Simonsen, et al. Physical review letters 100.21 (2008): 218701.)

Fragility model: initial failures

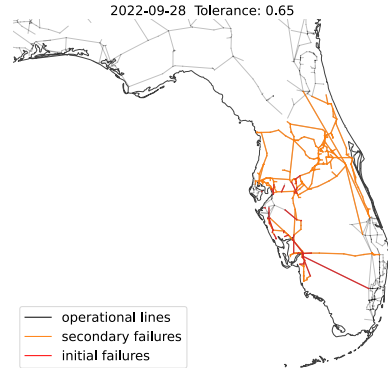
Link failure with probability p
and wind speed w (mph)

$$p(w) = \begin{cases} 0, & w < 30 \\ (w - 30)/30 & 30 \leq w < 60 \\ 1 & w \geq 60 \end{cases}$$



(Mathaios et al. 2017. "Power System Resilience to Extreme Weather: Fragility Modeling, Probabilistic Impact Assessment, and Adaptation Measures." IEEE Transactions on Power Systems 32 (5): 3747–57.)

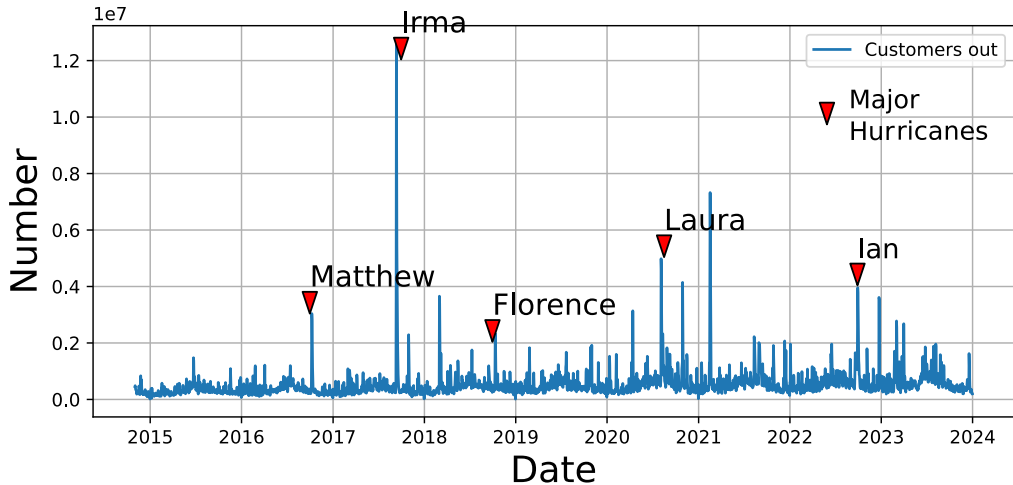
Hurricane hits Florida (2022)



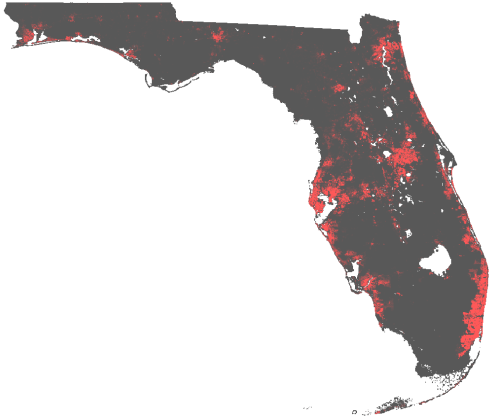
Wind strength

Simulated cascade

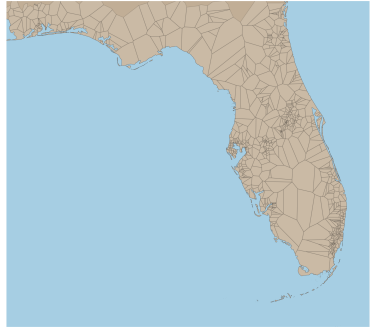
Blackout time series



Estimation of customers out

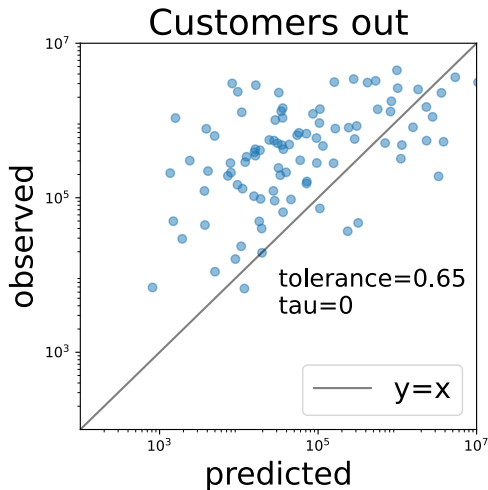


Population density



Voronoi tessellation of Florida grid

Best fit



- Each dot represents an extreme event in our dataset
- We find significant ($p \ll 0.01$) Spearman correlation between predictions and observations
- Best fitted parameter are $\alpha = 0.65$ and $\tau = 0$

Questions?

Joint work with:



Manlio De Domenico



Mauro Faccin

- Model of flow propagation in power grids
- Use of weather data to fit the model
- Use of historical data for validation



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Thanks for your attention!