

Avalanches in Complex Systems: From Sandpiles to Internet Storms

**Romualdo Pastor Satorras
Universitat Politècnica de Catalunya
Barcelona, Spain**

Summary

- Avalanche behavior in nature
- Explaining avalanches: the Self-Organized Criticality (SOC) concept
- A SOC example: sandpiles
- Boundary conditions: Sandpiles are not SOC ...
- Avalanches without SOC ?
 - Two not so typical examples
 - Turbidite deposition
 - Internet storms
- Conclusions

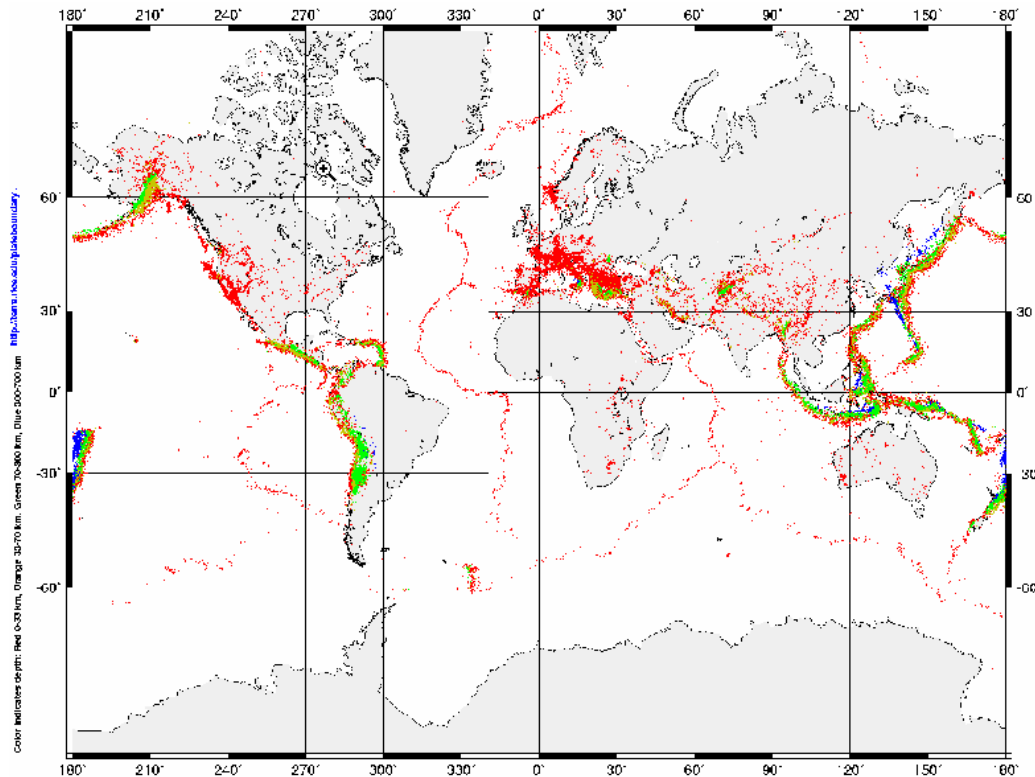
Avalanches in nature

- The activity of many complex systems, composed by many interacting units, is characterized by an avalanche behavior
 - Long times of quiescence, interspersed by bursts of sudden and very strong activity, that last a relatively short time
- Typical characteristic of avalanche activity
 - Avalanche size distributed according to a power law

$$P(s) \sim s^{-\tau}$$

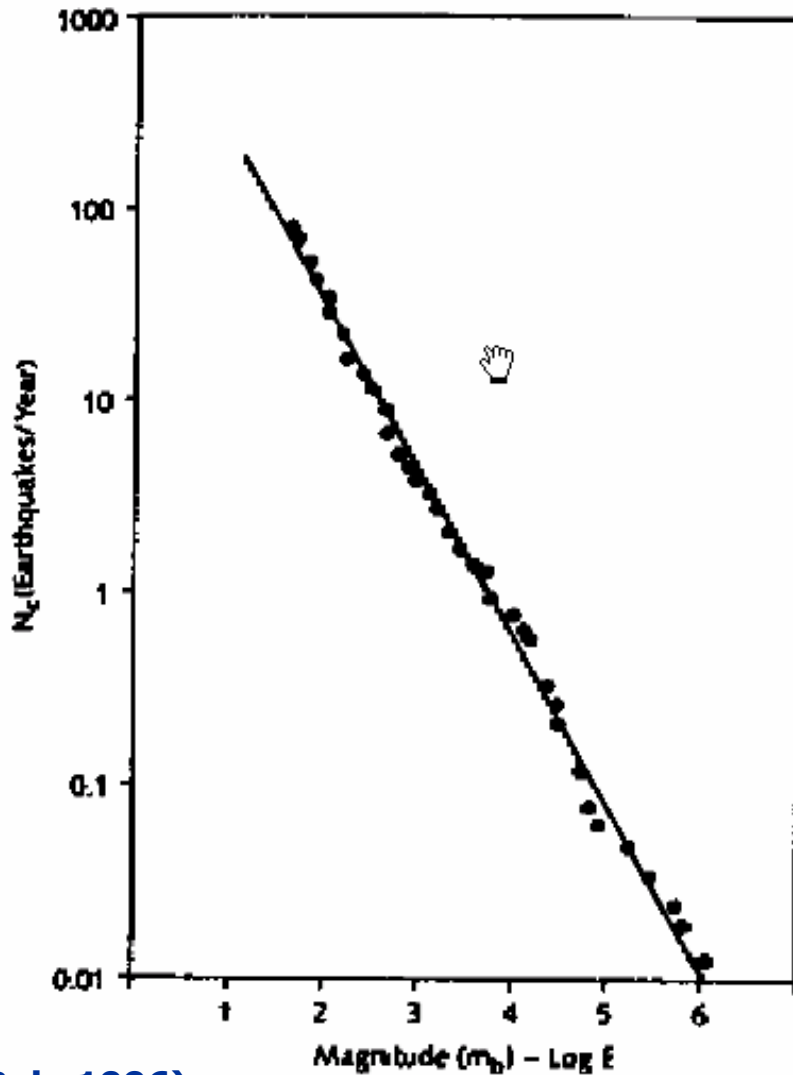
- Power law activity has been associated with natural catastrophes
 - Many physical systems show however this kind of behavior
- Examples ...

Earthquakes



- Earthquakes typically result from the movement of geological faults
- After long time accumulating stress, it is released in a sudden (and catastrophic) way

Earthquakes



(Bak, 1996)

KITP, Santa Barbara, 2005

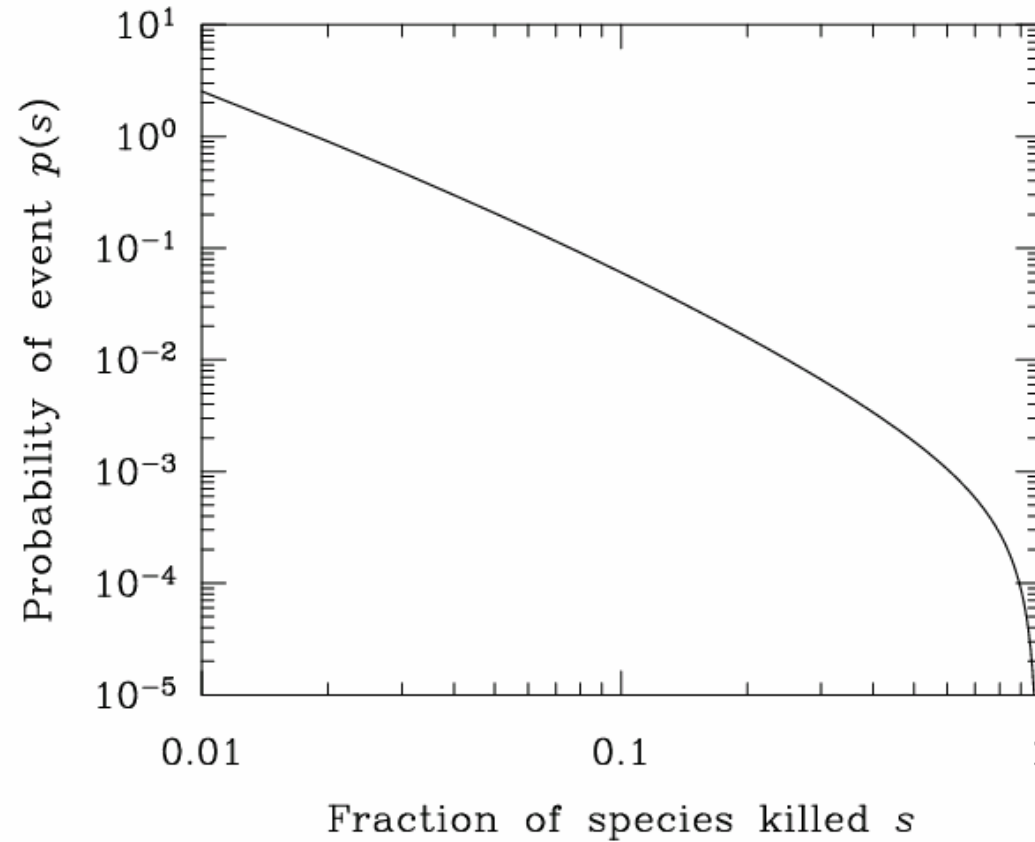
- Gutenberg-Richter law
 - The probability of observing an earthquake of a given size (magnitude) decreases as a power law

Extinctions



- Extinction of species takes place in short episodes (extinction events) that can involve a large number of species
- Some extinction events have been explained as real catastrophes (Alvarez et al., 1980)
- Many extinctions cannot be explained this way
 - Competition stress between species

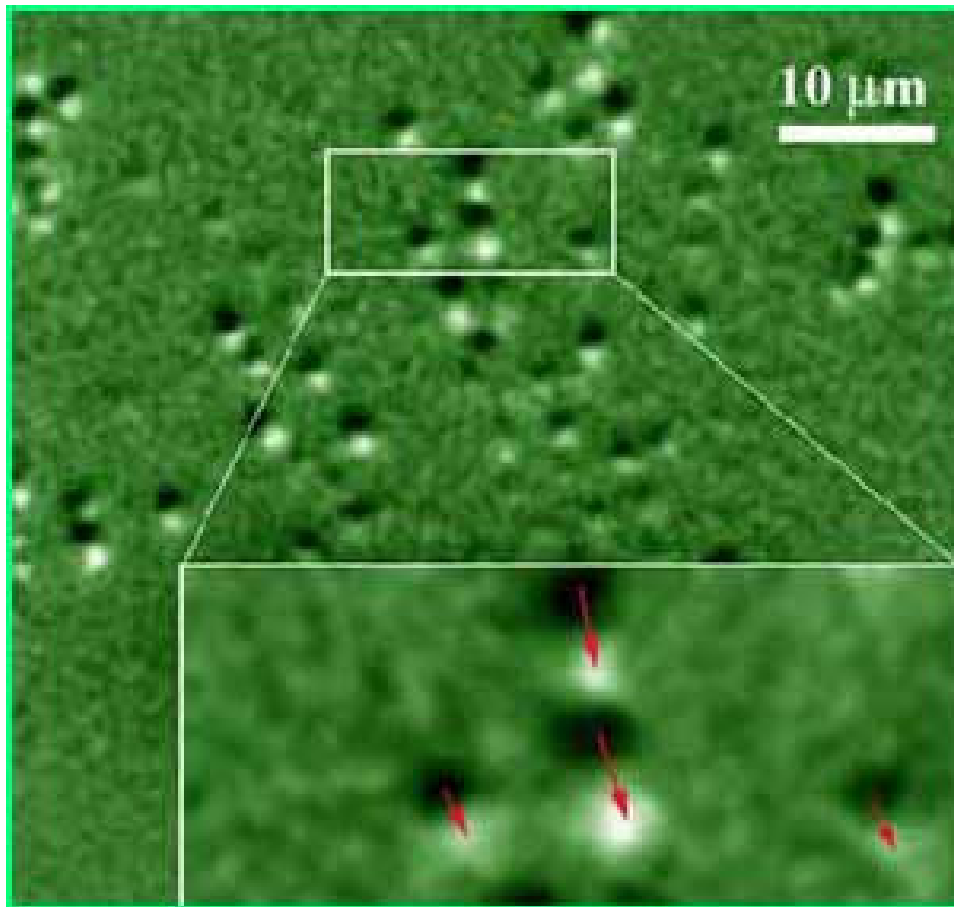
Extinctions



(Newman, 1997)

- The distribution of the size (fraction of species involved) in extinction events is approximately given by a power law

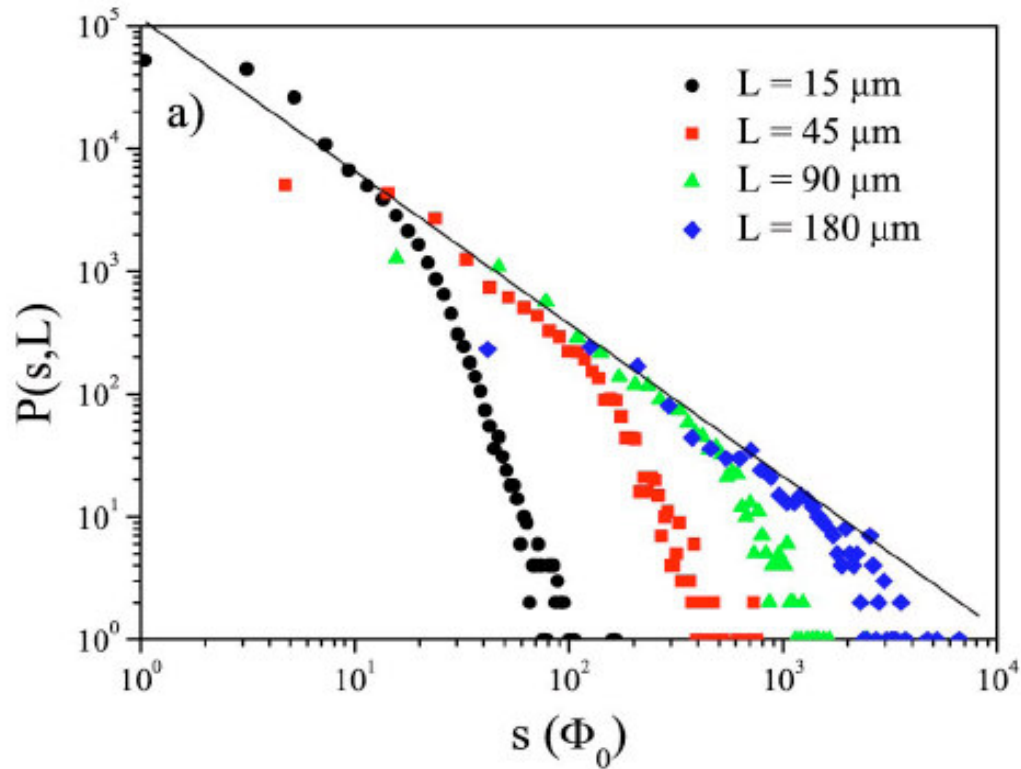
Superconductor vortices



- When a type II superconductor is driven by a slowly increasing magnetic field, vortex lines move inside the sample in a collective avalanche-like way
- This motion is due to the interplay among vortex interactions, quenched disorder, and field driving

(Altshuler et al., 2004)

Superconductor vortices



(Aegerter et al., 2003)

- The size of the avalanches, measured as the number of vortices involved in the movement, scales as a power law

Other avalanche examples

- Physics
 - Dislocation motion
 - Barkhausen effect
 - Charge density waves ...

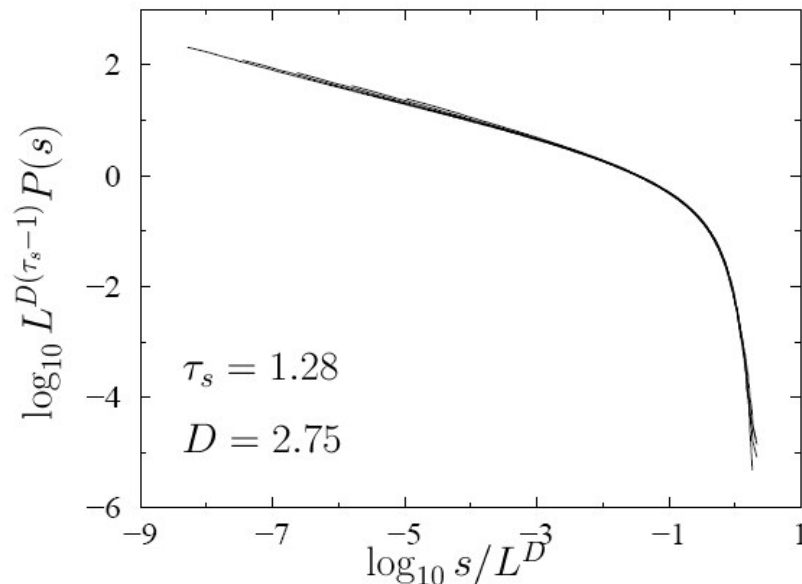
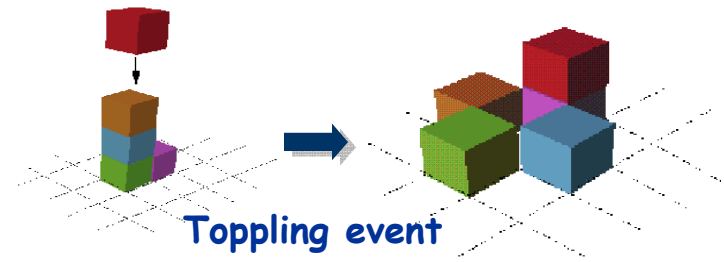
- Complex systems
 - Forest fires
 - Landslides
 - Traffic jams
 - Snow avalanches ...

An explanation for avalanche behavior?

- The wide observation of avalanche behavior with a power law distribution calls for some mechanism to explain it
- Per Bak et al. argument (1987):
 - Power law behavior is common in standard critical points
 - Systems with avalanches could be placed at a critical point?
 - Standard critical points are reached by adjusting tuning parameters ...
- Is it possible to have systems that adjust themselves, without any external control, to a critical point?
- The existence of such systems can explain the presence of power laws in nature
- Concept of Self-Organized Criticality (SOC) !!

An example of SOC

- Bak-Tang-Wiesenfeld sandpile model
 - Grains of sand (energy) are injected on a lattice
 - Conserved threshold dynamics of energy
 - Energy is lost at the boundary
 - The addition of a grain of energy can lead to an avalanche of activity



- The size of the avalanches (number of toppling events) scales as a power law.
- Moreover, it shows critical behavior: finite-size scaling

$$P(s) = s^{-\tau} f(s/L^D)$$

A closer look at boundary conditions ...

- So sandpile models seem to exhibit SOC (we have tuned no parameter)
- Let's have a closer look at the boundary conditions ...
 - Driving: We are adding energy when there is no activity (no topplings)
 - Dissipation: Energy dissipates at the boundary when there is activity
- Look at the energy E in the system ...
 - If E is large, there will be activity
 - If E is small, there will be no activity
- Balance of the energy ...
 - E large \rightarrow There is dissipation $\rightarrow dE/dt < 0$
 - E small \rightarrow There is driving $\rightarrow dE/dt > 0$
- Boundary conditions are driving the system to a particular value of the energy E_c ,
- If this energy corresponds to a critical point:
 - Sandpiles would have a built-in "baby-sitter" driving them to a conventional critical point (Dickman et al. 2000) !!

Firing the baby-sitter ...

- Consider a sandpile model in which we have fired the baby-sitter
 - No driving (no particles added)
 - No dissipation (periodic boundary conditions)
- Fixed energy sandpile (FES)
- Control parameter: energy of the system E
 - E small \rightarrow no sites above threshold \rightarrow the system is frozen
 - Absorbing phase
 - E large \rightarrow some sites above threshold \rightarrow the system is active
 - Active phase
- The transition between active and absorbing phases is a standard dynamic critical point (absorbing-state phase transition) taking place at a critical value of the energy E_c

Absorbing-state phase transition in FES

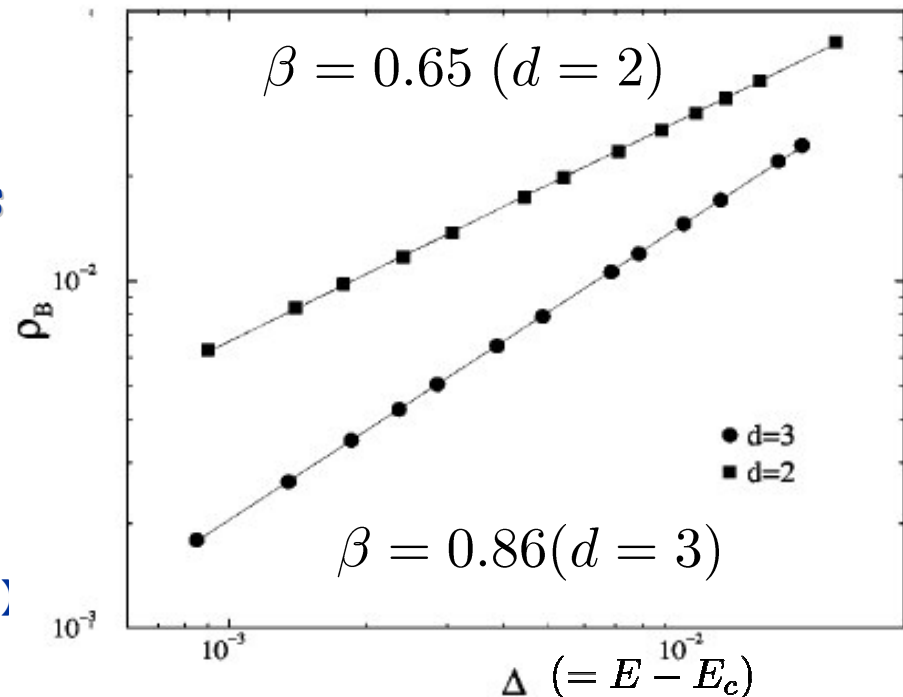
- General absorbing-state phase transitions can be studied by the methods used for general critical points
 - Definition in terms of a set of critical exponents ($\beta, \nu_{\perp}, \nu_{\parallel}$)

$$\rho \sim (E - E_c)^{\beta}, \quad \xi \sim (E - E_c)^{-\nu_{\perp}}, \quad \tau \sim (E - E_c)^{-\nu_{\parallel}}$$

- Computer simulations show that indeed FES undergo a completely standard absorbing-state phase, characterized by a set of exponents that define the universality class of sandpiles

(RPS & Vespignani, 2000)

KITP, Santa Barbara, 2005



Sandpiles are not self-organized ...

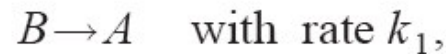
- Further observations:
 - The critical energy of the FES E_c and the average energy of the sandpile coincide
 - The sandpile critical exponents (τ, D) can be related to the FES exponents $(\beta, v_{\perp}, v_{\parallel})$
- Conclusion: Sandpiles are the other side of the coin of a standard absorbing-state transition critical point
 - Instead of adjusting the value of the control parameter, we adjust the boundary conditions so that the system is driven to the critical point
- From a more technical point of view:
 - Sandpiles are spreading experiments in FES
 - Measure of activity created by a single active site at the critical point

Field-theory for sandpiles

- If sandpiles are systems with an absorbing-state phase transition, it should be possible to characterize them by a field-theory (Reggeon field-theory for directed percolation)
- Too difficult for sandpiles (threshold dynamics)
- Universality conjecture (Rossi, RPS, & Vespignani, 2000):
 - All systems with the same symmetries as sandpiles belong to the same universality class
- Checking the conjecture with non-sandpile models
 - Conserved lattice gas (Rossi, RPS & Vespignani, 2000)
 - Reaction-diffusion system (RPS & Vespignani, 2000)
- Look for a model in the same universality class that is easier to treat analytically

Field-theory for sandpiles

- Consider a reaction-diffusion process with the same symmetries (in the same universality class)



- A field-theory can be constructed using standard formalism (Doi, Peliti, ...):

$$\partial_t \psi = D \nabla^2 \psi - r \psi - u_1 \psi^2 - u_2 \psi \phi + \eta_\psi,$$

$$\partial_t \phi = \lambda \nabla^2 \psi + \eta_\phi.$$

$$\langle \eta_\psi(x, t) \eta_\psi(x', t') \rangle = 2u_1 \psi(x, t) \delta(x - x') \delta(t - t')$$

$$\langle \eta_\psi(x, t) \eta_\phi(x', t') \rangle = -u_2 \psi(x, t) \delta(x - x') \delta(t - t')$$

$$\langle \eta_\phi(x, t) \eta_\phi(x', t') \rangle = 0$$

(RPS & Vespignani, 2000)

Can we explain avalanches without SOC?

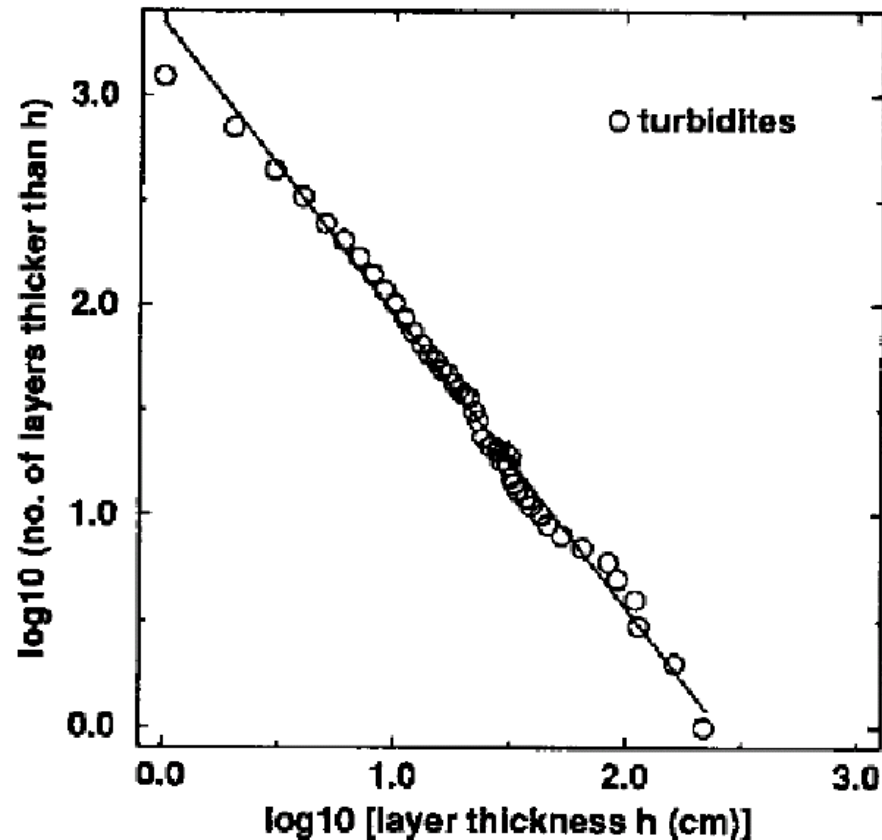
- SOC (or at least sandpiles) does not seem to be a robust paradigm to explain in general the wide presence of power law avalanche behavior in complex systems
 - They are implicitly tuned to a standard critical point by boundary conditions
- Can we say something otherwise about the avalanche behavior observed in nature?
- Answer: Yes, at least in some particular systems
- Examples:
 - Turbidite deposition
 - Internet blackouts (storms)

Turbidite deposition

- Sloping topographies erode in infrequent events (avalanches)
- In submarine topographies, avalanches create gravity-driven flows on material that, after sedimentation, form sedimentary rocks (turbidites)
- Turbidites are formed by different layers, corresponding to different avalanches
- Thickness of layers is related to the size of avalanches



Turbidite deposition



- The thickness distribution of turbidite layers follows in some cases a power law distribution

$$P(\Delta) \sim \Delta^{-\gamma}$$

- Characteristic exponent

$$\gamma \sim 2 - 2.4$$

One step back ...

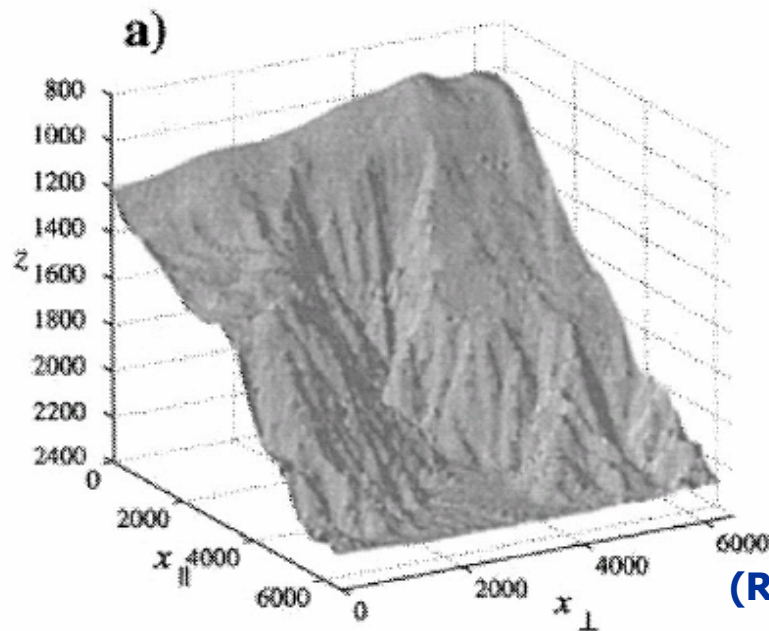
- To understand the scaling of turbidite deposition we must go one step back: consider the topography that generates the avalanches
- Topographical surfaces are rough
 - Described in terms of self-affine surfaces
 - Characterized by means of the height-height correlation function

$$C(r) = \langle |h(x+r) - h(x)|^2 \rangle_x^{1/2} \sim r^\alpha$$

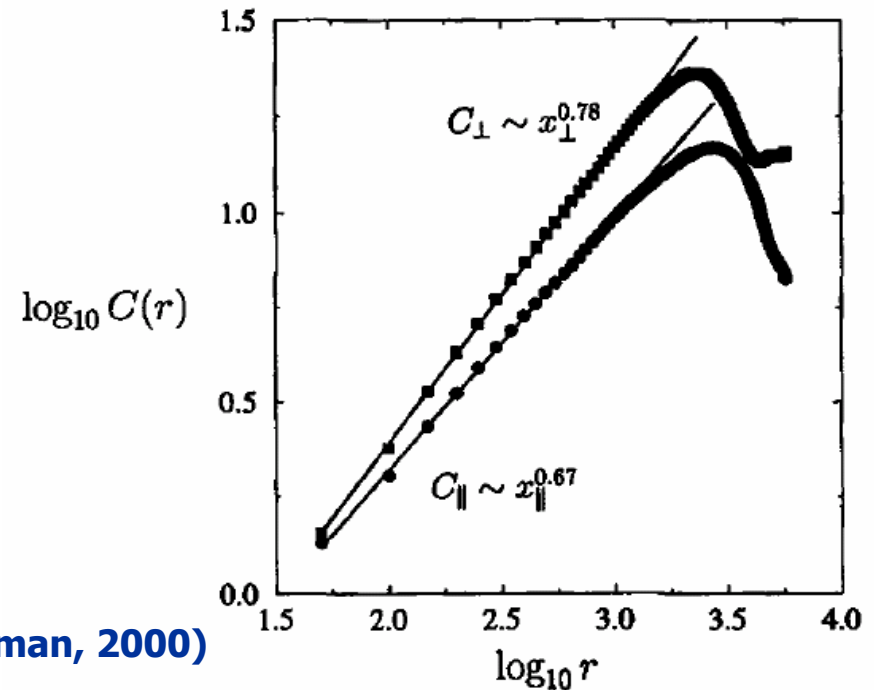
- α : characteristic roughness exponent
- Measured values ranging from 0.4 to 0.8 over the world

Tilted topographies

- Many topographies show in addition a average slope
 - Induces a preferred direction (downwards)
- This anisotropy generally induces the presence of two roughness exponents, when measuring correlation in the downwards direction or x_{\parallel} or in the perpendicular direction x_{\perp}



(RPS & Rothman, 2000)



A theory for tilted topographies

- Tilted topographies can be analytically studied using the techniques of self-affine growing surfaces:
- Construction of a stochastic growth equation for the landscape height $h(x,t)$, following symmetry arguments:
 - Anisotropy
 - Preferred transport along downwards x_{\parallel} direction
 - Conservation of material
- Lowest order equation

$$\frac{\partial h}{\partial t} = v_{\parallel} \partial_{\parallel}^2 h + v_{\perp} \nabla_{\perp}^2 h + \frac{\lambda}{3} \partial_{\parallel}^2 (h^3) + \eta$$

Source of
random noise



(RPS & Rothman, 1998)

RG analysis

- Applying the dynamic renormalization group in a one-loop ε -expansion

$$\alpha_{\perp} = \frac{5\varepsilon}{12}, \quad \zeta_{\perp} = 1 + \frac{\varepsilon}{6}, \quad \varepsilon = 4 - d \quad \alpha_{\parallel} = \frac{\alpha_{\perp}}{\zeta_{\perp}}$$

(RPS & Rothman, 1998)

- At the physical dimension $d=2$

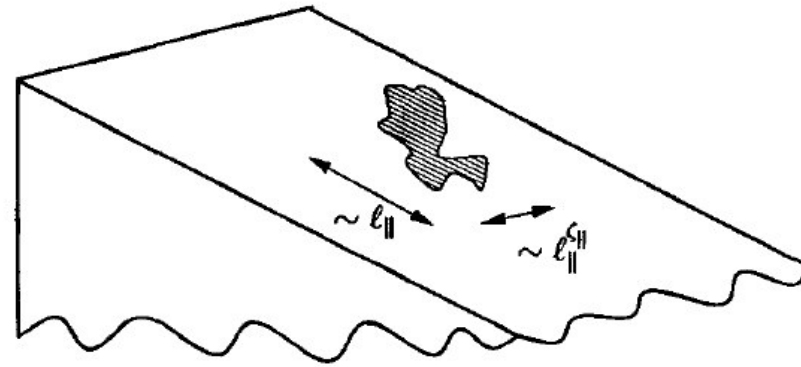
$$\alpha_{\perp} = \frac{5}{6} \simeq 0.83, \quad \zeta_{\perp} = \frac{1}{\zeta_{\parallel}} = \frac{4}{3}, \quad \alpha_{\parallel} = \frac{\alpha_{\perp}}{\zeta_{\perp}} = \frac{5}{8} \simeq 0.63$$

- Reasonable agreement with field measurements ($\alpha_{\parallel} = 0.67$, $\alpha_{\perp} = 0.78$)

Back to turbidites ...

- Assume turbidite layers come from unstable patches of terrain that fall from a tilted submarine landscape
- Size of patches scales as the surface

$$s \sim l_{\parallel} l_{\perp} \sim l_{\parallel}^{1+\zeta_{\parallel}}$$



- Assuming a power law distribution of patches sizes

$$P(s) = s^{-\tau} f\left(\frac{s}{L^{1+\zeta_{\parallel}}}\right)$$

Back to turbidites ...

- Imposing the condition $\langle s \rangle \sim L$ (patches are very elongated)

$$\tau = 2 - \frac{1}{1 + \zeta_{\parallel}}$$

- Relating thickness with size $\Delta \sim s^{1/3}$ we obtain the thickness distribution

$$P(\Delta) \sim \Delta^{-\gamma} \quad \gamma = 1 + \frac{3\zeta_{\parallel}}{1 + \zeta_{\parallel}}$$

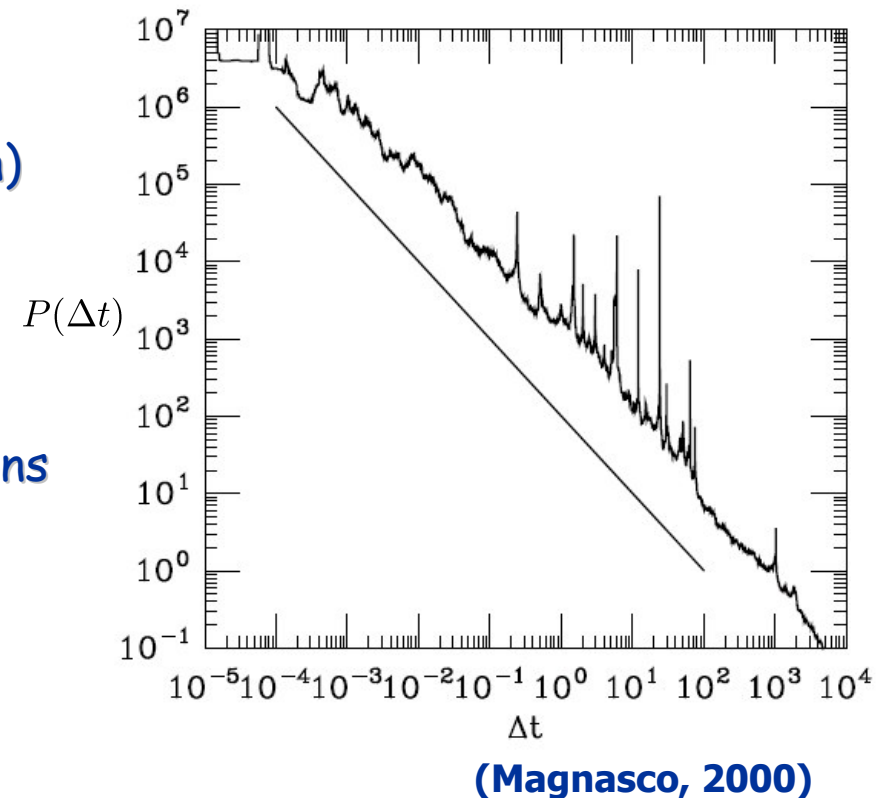
- Inserting values from RG: $\gamma \approx 2.3$
 - Good agreement with field values

Internet storms: Instabilities and congestions

- Everybody has experienced short-term Internet outages:
 - Clicking on a link that does not respond, but responds a few seconds later
- Internet outages usually come from instabilities propagating through the network
 - Configuration errors
 - Traffic congestions
 - Software bugs ...
- The propagation of instabilities and congestions can lead to failure avalanches that can collapse regions of the Internet, blocking its access

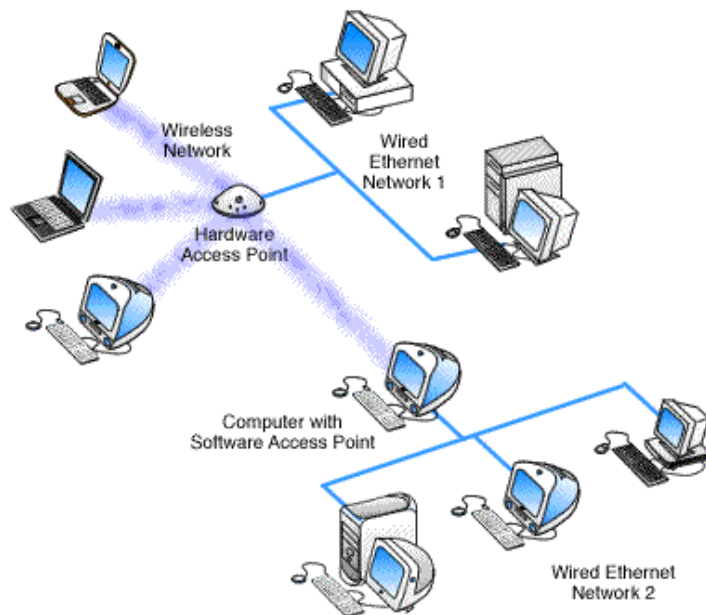
Internet storms: Instabilities and congestions

- Empirical measurements show that Internet congestions have a power law signatures
 - Distribution of interarrival error messages (signature of congestion) are distributed according to a power law of exponent -1
 - Measures of the size of congestions are more difficult but seem to hint towards the same power law trend



One step back ...

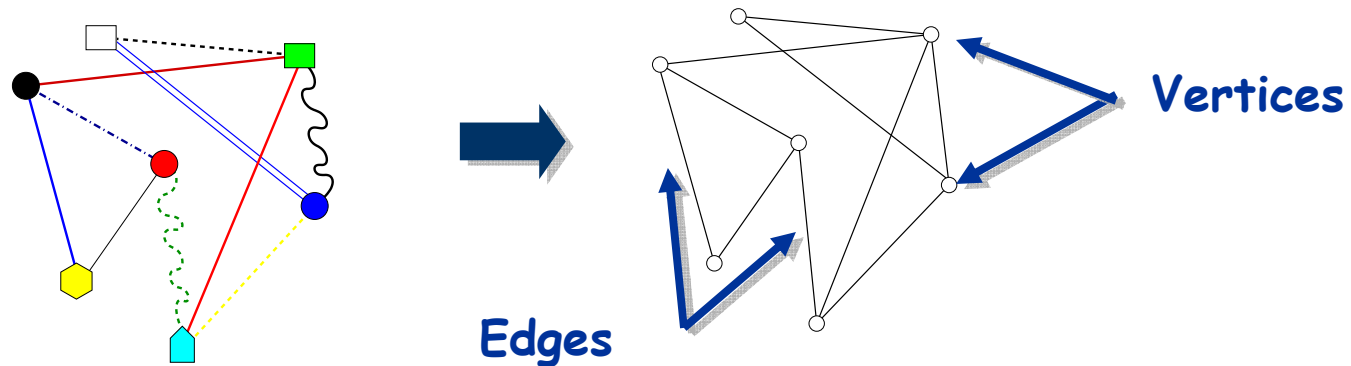
- We can understand Internet storms by understanding the Internet topology
- Key concept: Computer network
 - Set of interconnected computers that can communicate among each other



- Internet: network that interconnects many different computer networks on a world-wide scale
- Main characteristic:
 - Heterogeneity both in design and components

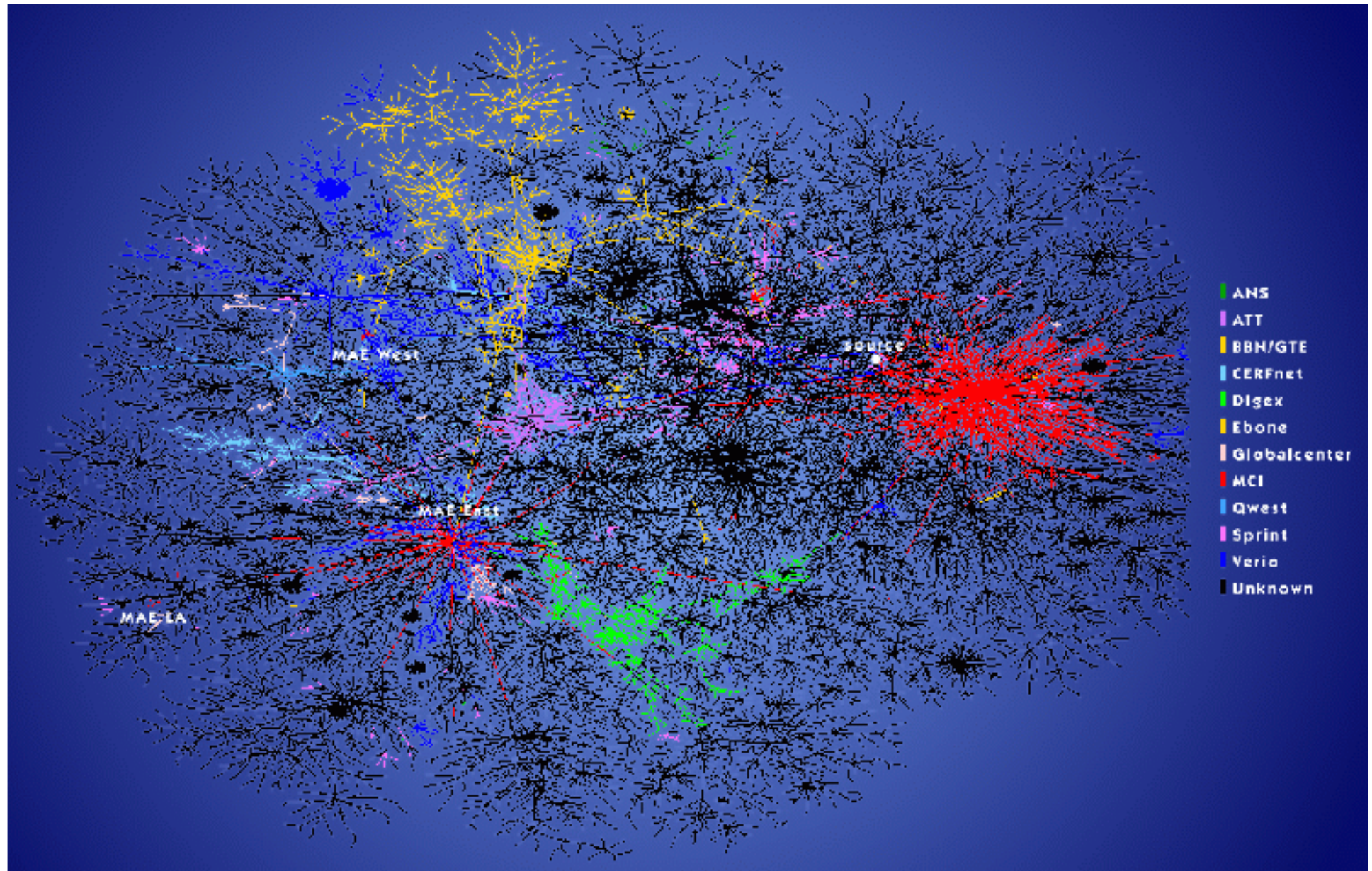
Topology of the Internet

- The Internet belongs to a general class of complex heterogeneous systems
 - Large number of diverse elementary components
 - Interactions between components can also diverse and can be non-local
 - Number of elements $10^3 - 10^6$
- New perspective for the study: representation as a graph or network



- Study of the topological properties of the representative network

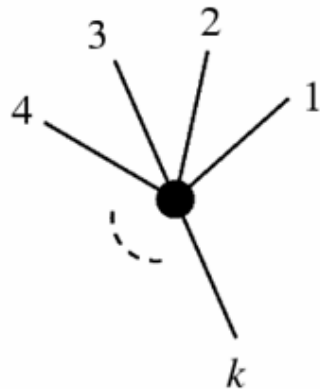
Representations of the Internet



Topological analysis of networks

- The most important topological characterization of a network is the degree of the vertices

- Degree k_i



- For large networks: Statistical characterization by the degree distribution

$$P(k)$$

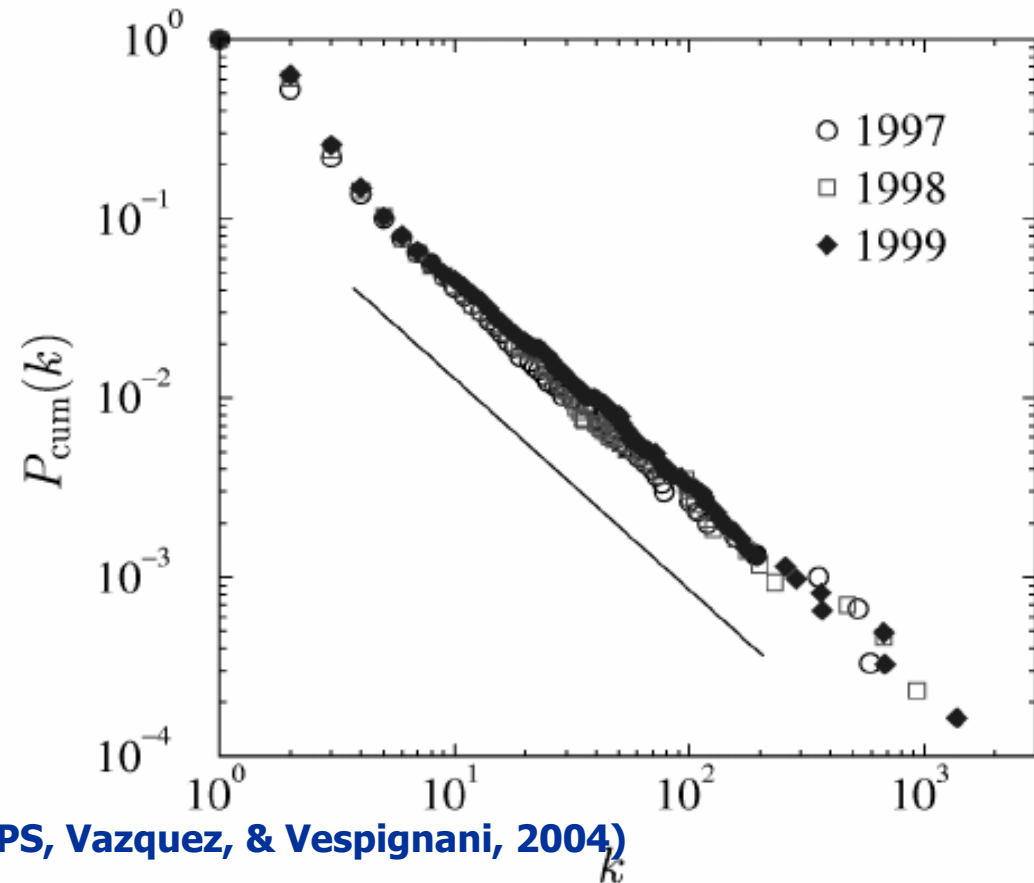
Scale-free Internet

- Empirical investigation shows that the Internet is a scale-free network
 - Degree distribution given by a power law

$$P(k) \sim k^{-\gamma}$$

- Degree exponent

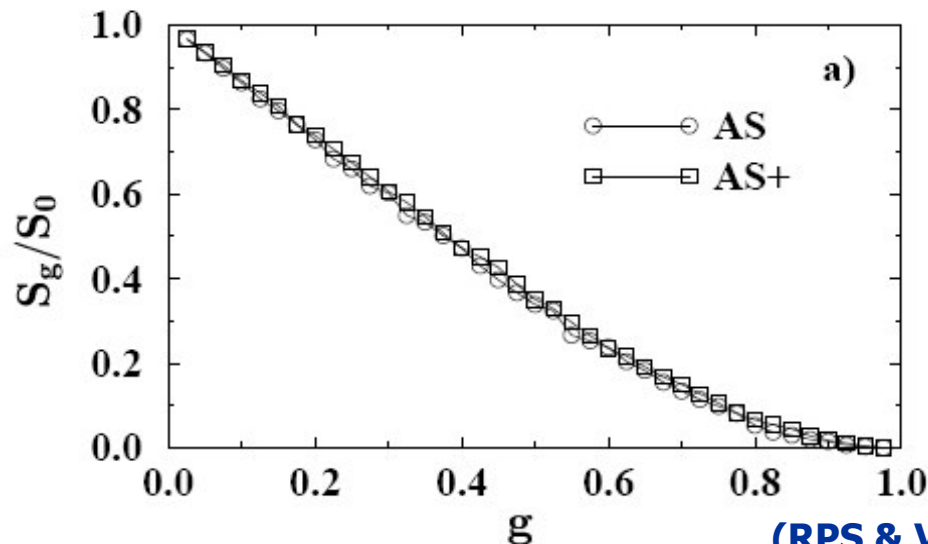
$$\gamma \sim 2.2$$



Effects of an scale-free topology

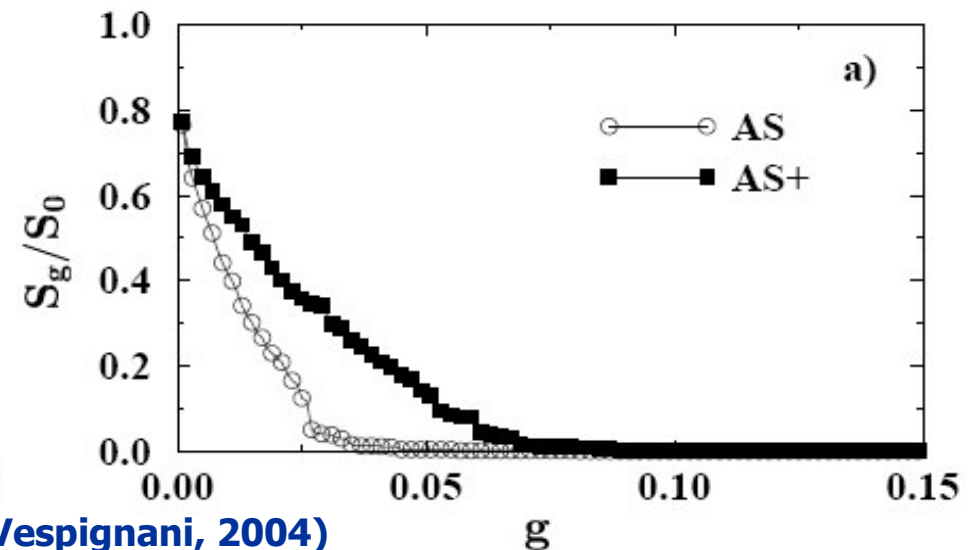
- A scale-free topology can have a very strong impact on the properties of the system
 - The Internet is very strong against random removal of elements
 - On the other hand, it is very weak against the targeted removal of the most connected elements

Random removal of elements



MIT, Santa Barbara, 2003

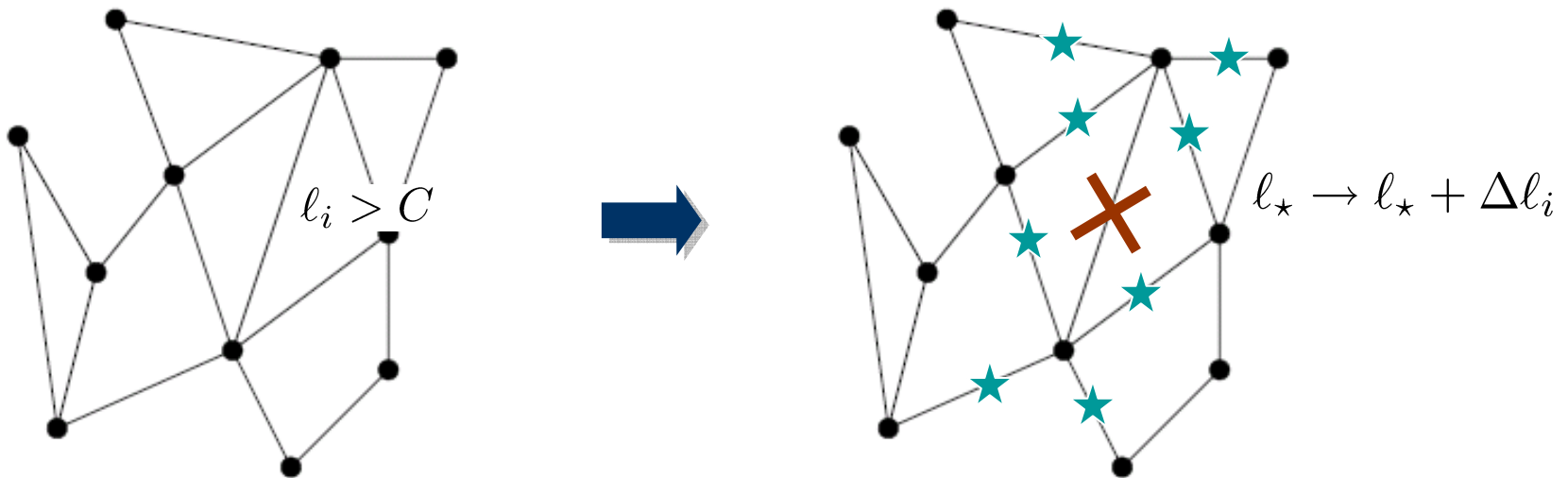
Targeted removal of elements



(RPS & Vespignani, 2004)

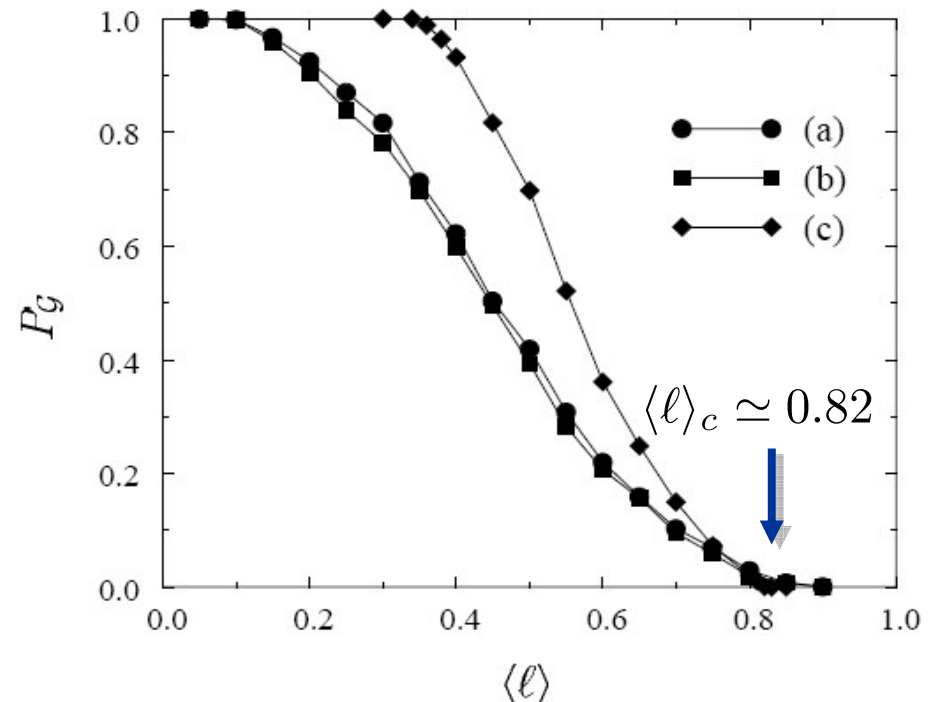
Back to Internet storms ...

- Can the topology of Internet explain the distribution of outages?
- Simple model of the static failure of a transport network
 - Scale-free network
 - Each connection (edge) carries load l_i at random from a uniform distribution with average $\langle l \rangle$
 - Threshold dynamics: when $l_i > C$, the connection breaks and its load is redistributed among nearest connections
 - The redistribution can produce an avalanche of connection failures

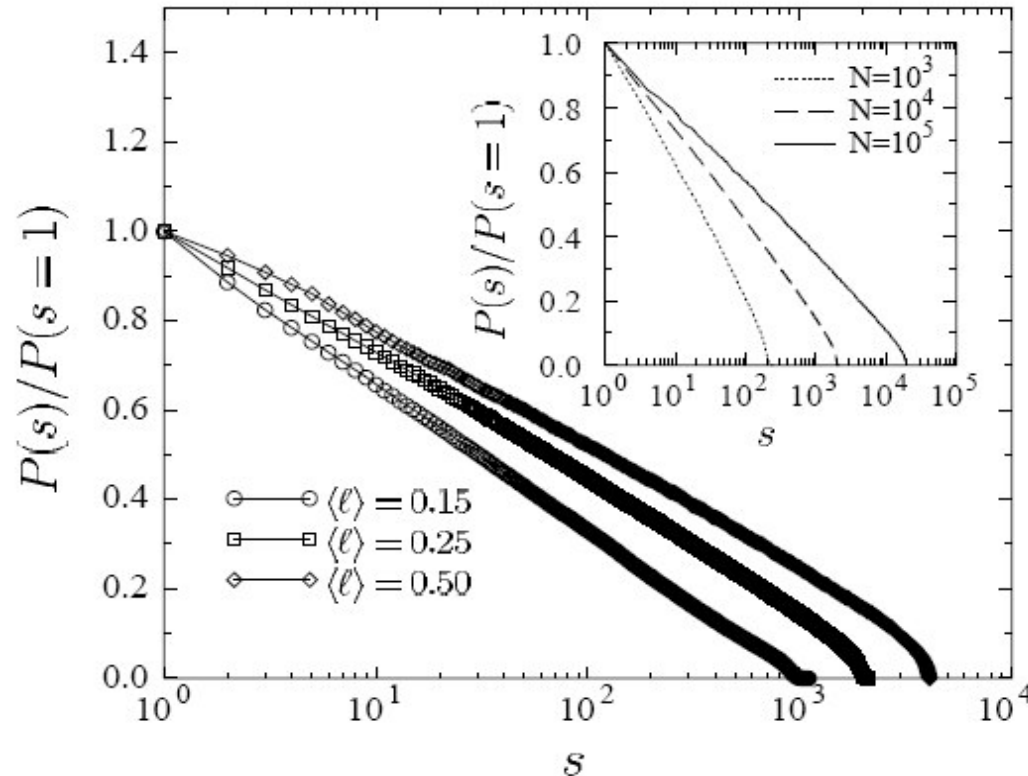


Back to Internet storms ...

- When increasing $\langle \ell \rangle$, connections break and the network becomes disconnected
- Study of the size of the largest piece of connected network as a function of the average load
 - There is a phase transition at a finite value $\langle \ell \rangle_c$, separating a disconnected network (with no communication capabilities) from a connected network
- Note: We need to adjust $\langle \ell \rangle_c$
 - Usual phase transition



Back to Internet storms ...



(Moreno, RPS & Vespignani, 2003)

- Numerical study of failure avalanches: distribution of broken connections
- Distribution size scales as a power law with exponent -1, for q wide range of values of $\langle \ell \rangle$
 - No need of fine-tuning
- Preliminary results on a dynamic model indicate distribution of times between avalanches as a power law with exponent -1

Conclusions

- Sandpile models are not a good paradigm for possible SOC
 - The boundary conditions drive implicitly the system towards a critical point
 - The critical point in sandpiles is a standard one (absorbing-state phase transition), where the tuning parameter is the energy density
- Without recurring to the SOC concept, we can still say things about avalanche behavior:
 - Avalanche properties can be connected to the geometrical properties of the system (tilted landscapes and turbidites)
 - The very topology of a system can lead to avalanche behavior without any fine-tuning (Internet storms)

Acknowledgements

- Collaborators:
 - Y. Moreno (Universidad de Zaragoza, Spain)
 - M. Rossi
 - D. H. Rothman (MIT, USA)
 - A. Vazquez (University of Notre Dame, USA)
 - A. Vespignani (Indiana University, USA)