

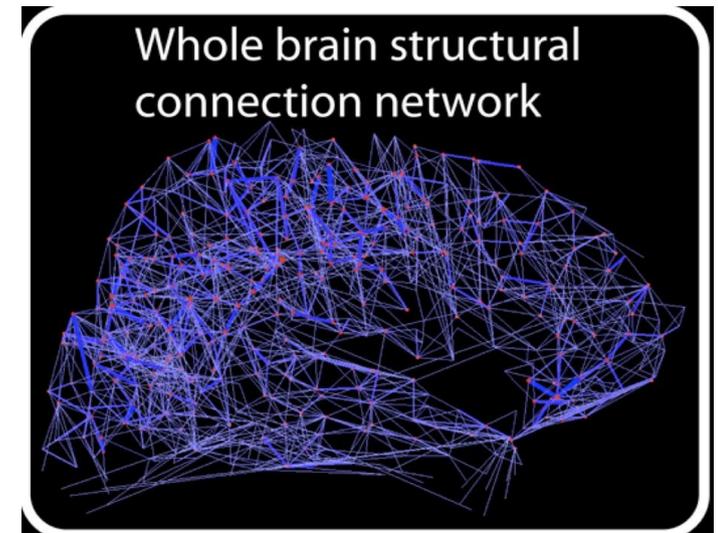
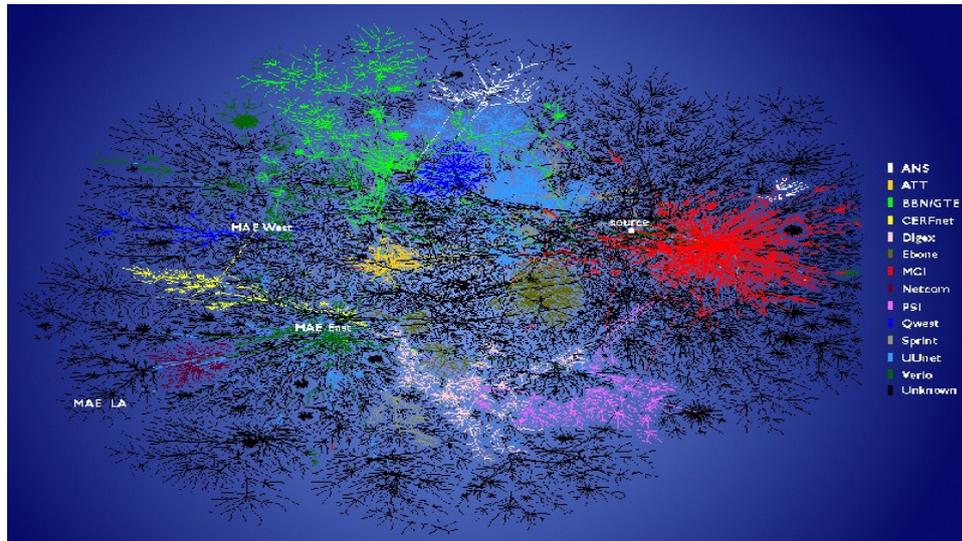
Slow dynamics of the contact processes on complex networks

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RESEARCH INSTITUTE FOR NATURAL SCIENCES (MFA) BUDAPEST

- *Exploration of complex networks is flourishing since ~2000 (Barabási & Albert)*
- *Dynamical systems living on networks are of current interest*
- *Origin of slow (dynamic) scaling behavior in internet, brain, quantum systems,... etc.*
- *Open question : Complex networks + quenched disorder ?*

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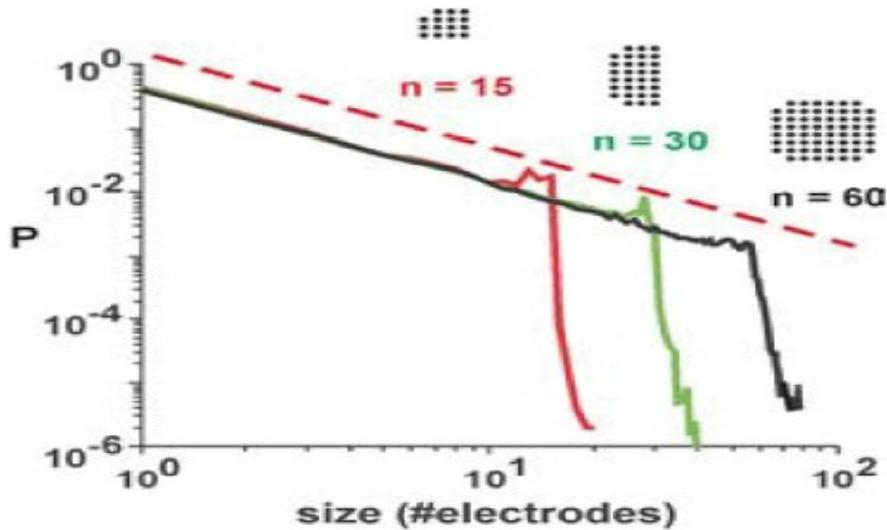


Diffusion spectrum imaging

Observed slow dynamics in networks

- Brain : Size distribution of neural avalanches

G. Werner : Biosystems, 90 (2007) 496,



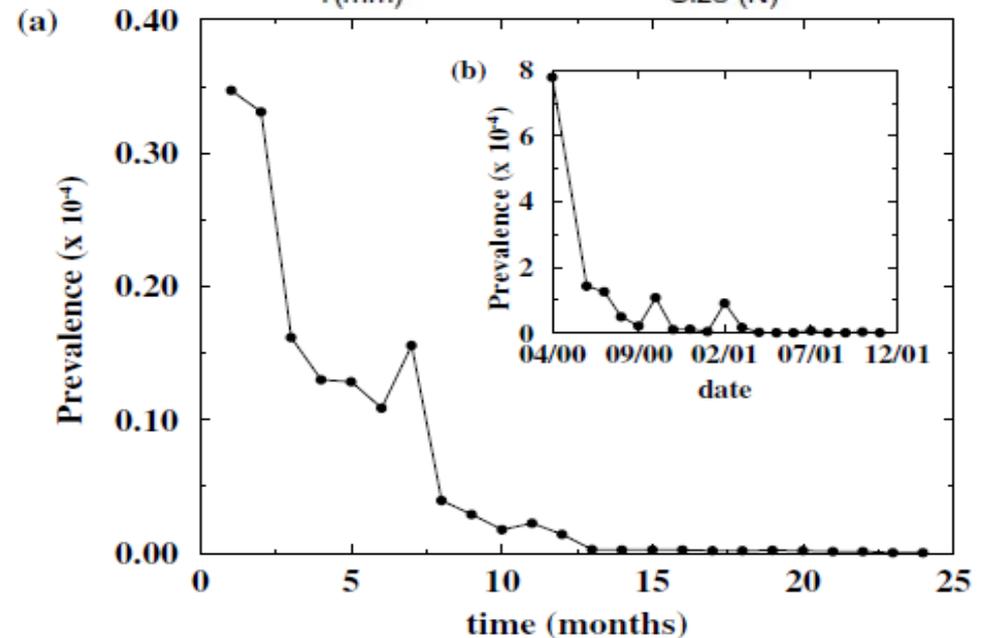
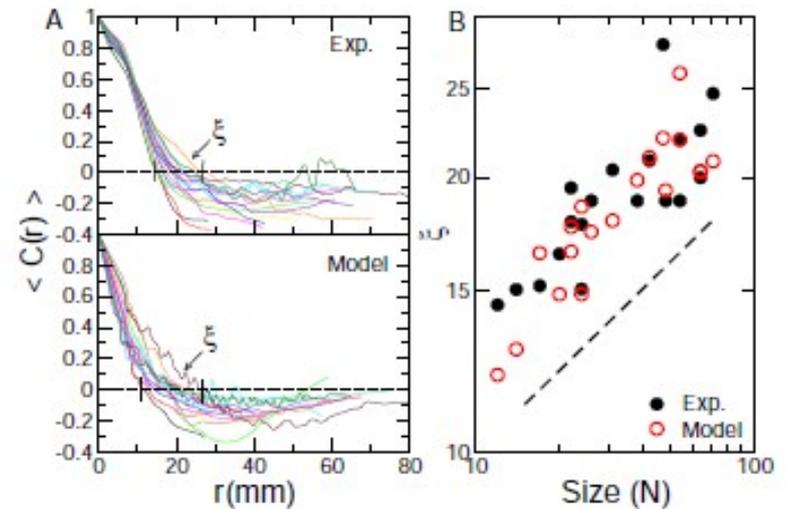
- Internet: worm recovery time is slow:

Small world networks \rightarrow fast dynamics

What is the cause ?

Correlation length (ξ) diverges

*Tagliazucchi & Chialvo (2012) :
Brain complexity born out of criticality.*



Slow dynamics, scaling in nonequilibrium

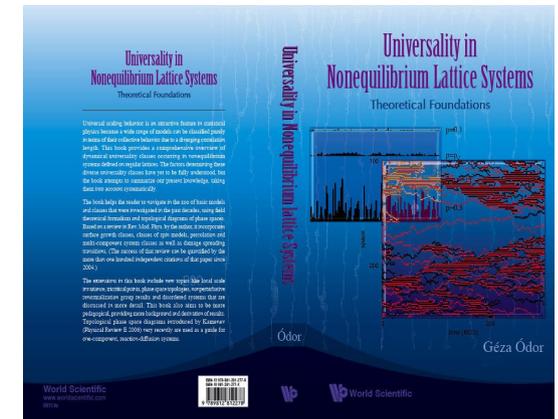
Scaling and universality classes appear in complex system due to : $\xi \rightarrow \infty$
i.e: near critical points

Basic models classified by universal scaling behavior

*G. Ódor: Universality in nonequilibrium system
(World Scientific 2008), Rev. Mod. Phys. 2004*

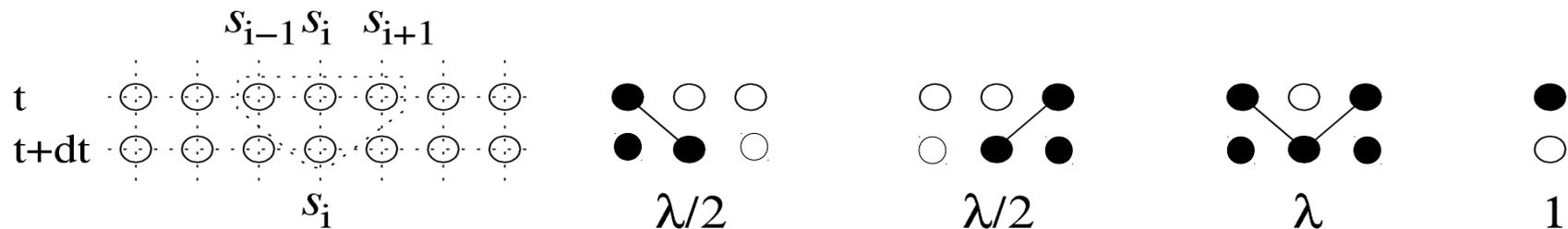
- **Why don't we see universal behavior in networks ?**
- **Tuning to critical point is needed !**

I'll show a possible way to understand this

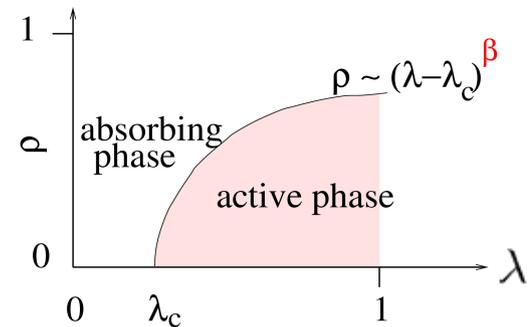


Modelling dynamics on fundamental (nonequilibrium) models

Prototype: **C**ontact **P**rocess describing “epidemic/info” propagation (1d) :



- In regular, Euclidean lattices:
 order parameter: ρ the density of active sites
 phase transition between active and inactive (absorbing)
 Critical point : $\lambda_c > 0$
- Exhibits scaling behavior belonging to the DP universality class, still rarely observed in nature
- Sensitivity to spatially/temporal (quenched) disorder → The scaling behavior is **slow, non-universal**



Rare Region argument for **Q-disordered CP**

- Fixed (quenched) disorder/impurity **changes the local birth rate** $\Rightarrow \lambda_c > \lambda_c^0$

- Locally active**, arbitrarily large Rare Regions

in the inactive phase due to the *inhomogeneities*

- Probability or RR of size L_R :

$$w(L_R) \sim \exp(-c L_R^d)$$

Contribute to the density: $\rho(t) \sim \int dL_R L_R^d w(L_R) \exp[-t/\tau(L_R^d)]$

- For $\lambda < \lambda_c^0$: conventional (exponentially fast) decay

- At λ_c^0 the characteristic time scales as: $\tau(L_R) \sim L_R^{-z} \Rightarrow$ saddle point analysis:

$$\ln \rho(t) \sim t^{d/(d+z)}$$

(stretched exponential)

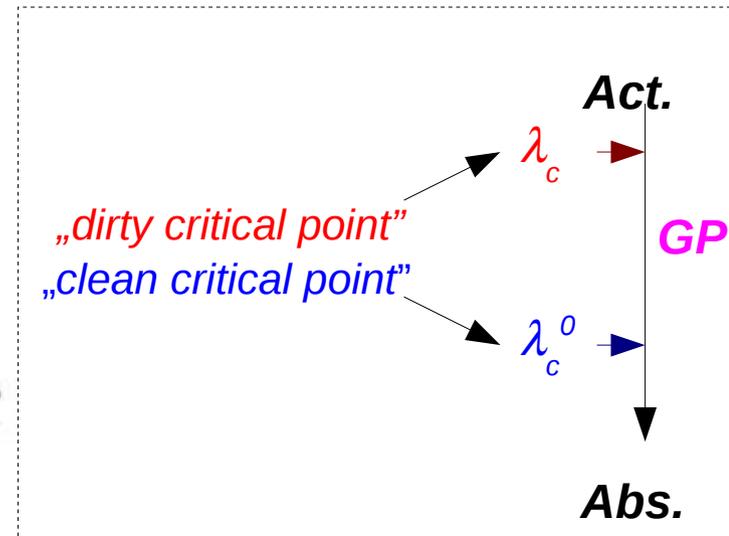
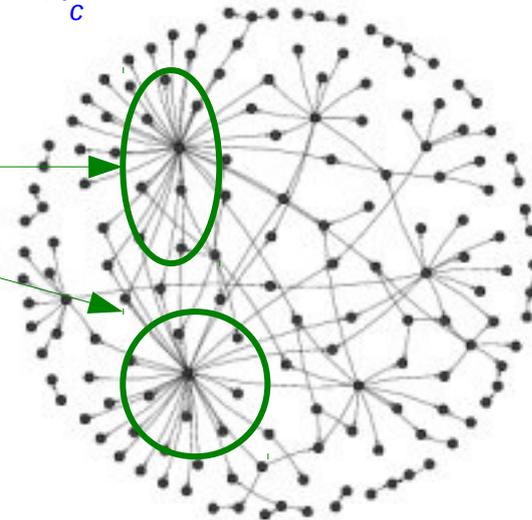
- For $\lambda_c^0 < \lambda < \lambda_c$: $\tau(L_R) \sim \exp(b L_R)$:

\Rightarrow saddle point analysis: $\rho(t) \sim t^{-c/b}$

Griffiths Phase

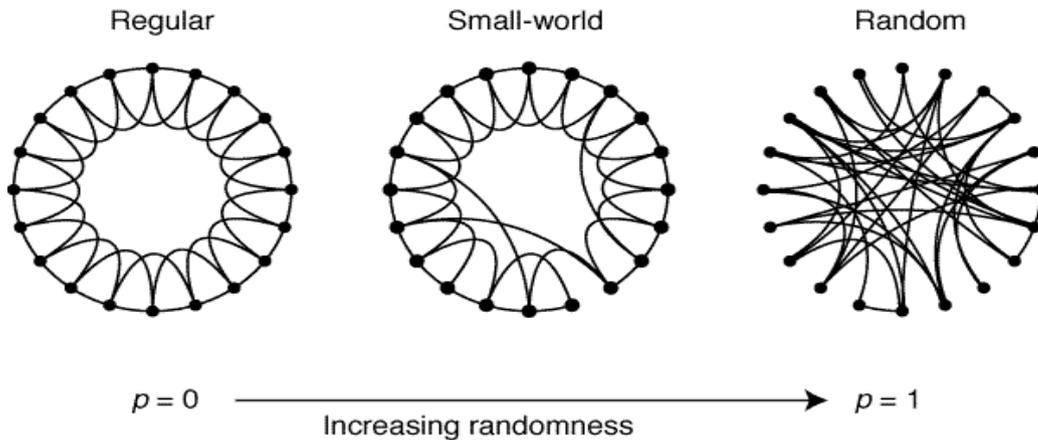
(continuously changing exponents)

- At λ_c Ultra slow time dependences: $\rho(t) \sim \ln(t)^{-\alpha}$



Basic network models

From regular to random networks:



Erdős-Rényi ($p = 1$)

Degree (k) distribution in $N \rightarrow \infty$ node limit:

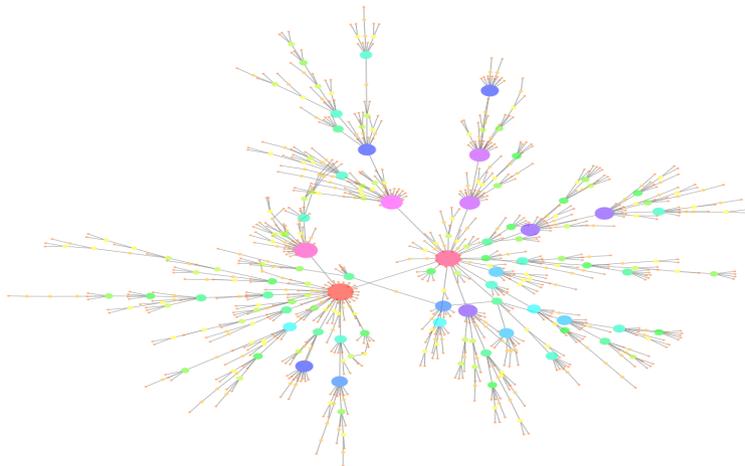
$$P(k) = e^{-\langle k \rangle} \frac{\langle k \rangle^k}{k!}$$

Topological dimension: $N(r) \sim r^d$

Above perc. thresh.: $d = \infty$

Below percolation $d = 0$

Scale free networks:



Degree distribution:

$$P(k) = k^{-\gamma} \quad (2 < \gamma < 3)$$

Topological dimension: $d = \infty$

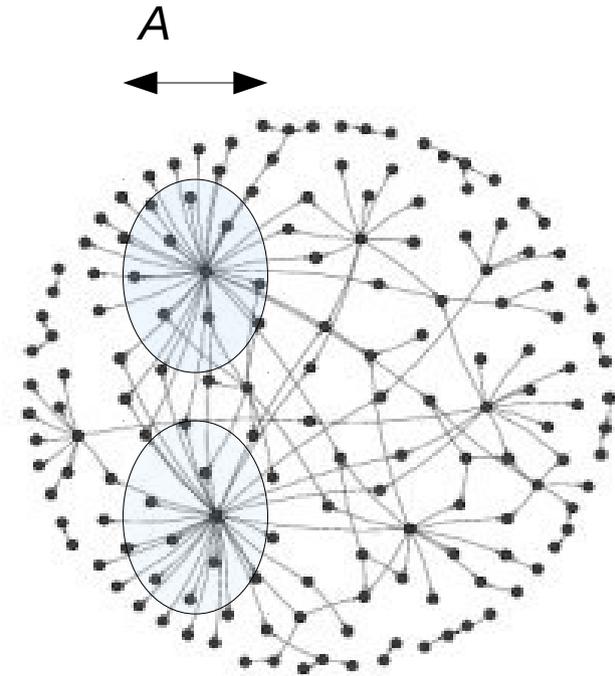
Example: **Barabási-Albert**
lin. preferential attachment

Focus on dynamical systems living on networks: Fast dynamics is expected

Networks: fast dynamics, mean-field behavior expected

Effect of disorder:

Rare active regions in the absorbing phase: $\tau(A) \sim e^A$
→ slow dynamics (Griffiths Phase) ?



M. A. Munoz, R. Juhász, C. Castellano and G. Ódor, PRL 105, 128701 (2010)

1. Inherent disorder in couplings
2. Disorder induced by topology

Optimal fluctuation theory + simulations:

- In Erdős-Rényi networks below the percolation threshold
- In generalized small-world networks for **finite topological dimension**

CP + Topological disorder results

Generalized Small World networks: $P(l) \sim \beta l^{-2}$
 (link length probability)

• Top. dim: $N(r) \sim r^d$ $d(\beta)$ **finite:**

$\lambda_c(\beta)$ decreases monotonically from

$\lambda_c(0) = 3.29785$ (1d **CP**) to:

$\lim_{\beta \rightarrow \infty} \lambda_c(\beta) = 1$ towards mean-field **CP** value

$\lambda < \lambda_c(\beta)$ inactive, there can be

locally ordered, rare regions due to more than average, active, incoming links

• **Griffiths phase:** λ -dep. continuously changing dynamical power laws:

for example: $\rho(t) \sim t^{-\alpha(\lambda)}$

Logarithmic corrections!

• **Ultra-slow** (“activated”) scaling: $\rho \propto \ln(t)^{-\alpha}$ at λ_c

• As $\beta \rightarrow 1$ Griffiths phase shrinks/disappears

• Same results for **cubic, regular random networks**
higher dimensions

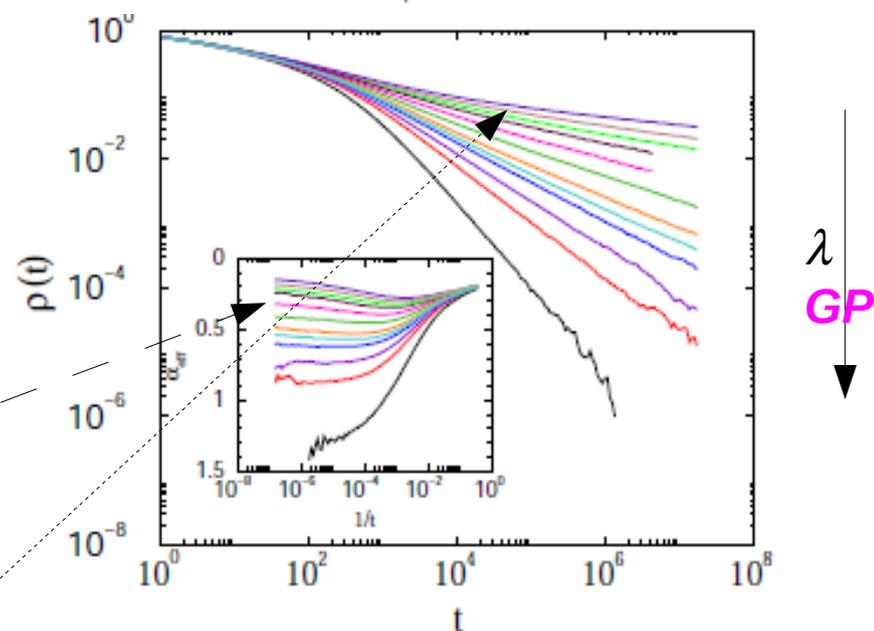
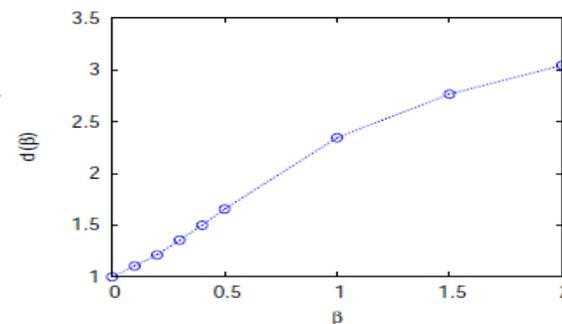
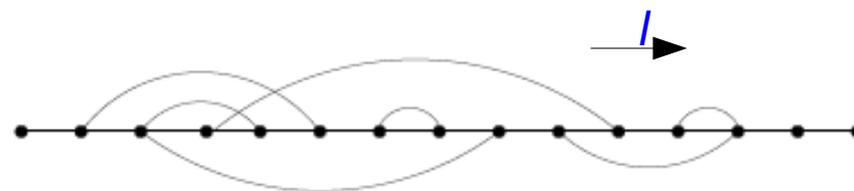


FIG. 3: Density decay in Benjamini-Berger networks with $s = 2$ and $\beta = 0.2$ for different values of λ (from top to bottom: 2.81, 2.795, 2.782, 2.77, 2.75, 2.73, 2.71, 2.70, 2.69, 2.67, 2.65, 2.6). Straight lines lie in the Griffiths phase. Inset: Corresponding effective exponents, illustrating the presence of corrections to scaling.

Contact process on Barabási-Albert (BA) network

- Heterogeneous mean-field theory: conventional critical point, with linear density decay:

$$\rho(t) \sim [t \ln(t)]^{-1},$$

with logarithmic correction

- Extensive simulations confirm this
- No Griffiths phase observed
- Steady state density vanishes at $\lambda_c \approx -1$ linearly, HMF: $\beta = 1$

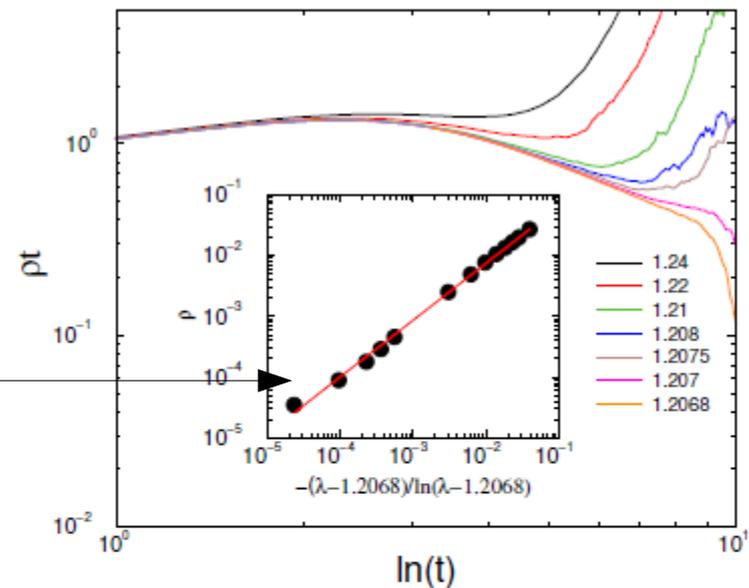


FIG. 1. Density decay ($t\rho(t)$) as a function of $\ln(t)$ for the CP on unweighted looped BA networks with $m = 3$ of size $N = 8 \times 10^7$. The different curves correspond to $\lambda = 1.2068, \dots, 1.24$ (bottom to top). Inset: Steady state density, showing agreement with HMF theory scaling. The full line shows a power-law fitting to the data points in the form $-0.36(5)x^{0.98(2)}$.

CP on Barabási-Albert trees

hunt for GP-s, by slowing the propagation

- Lack of loops slows propagation
- For $\langle k \rangle = 3 : \lambda_c > 0$

Strong size corrections

Non mean-field transition :

Weighted networks:

$$\omega_{ij} = \omega_0(k_i k_j)^{-\nu} \quad \omega_{ij} = \frac{|i-j|^x}{N}$$

$$k_i \propto (N/i)^{1/2}$$

Power-laws for: $x = 2, 3$

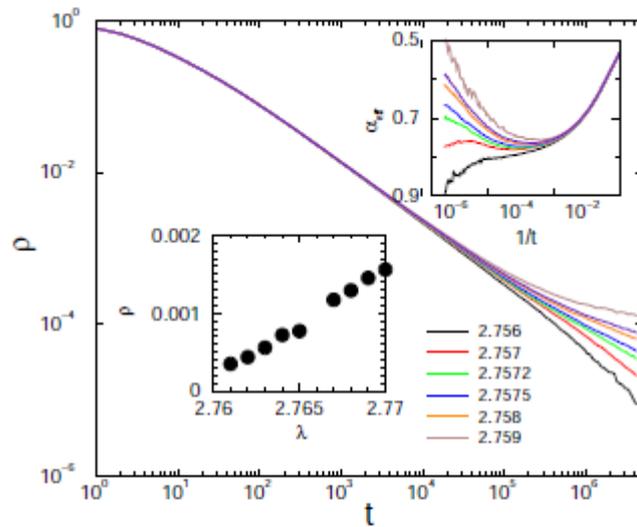


FIG. 1: Density decay in a pure BA CP model, $m = 1$, $m_0 = 10$, $N = 4 \times 10^7$ for $\lambda = 2.756, \dots, 2.759$ (bottom to top). Right insert: the corresponding effective exponents. Left insert: steady state density in the active phase.

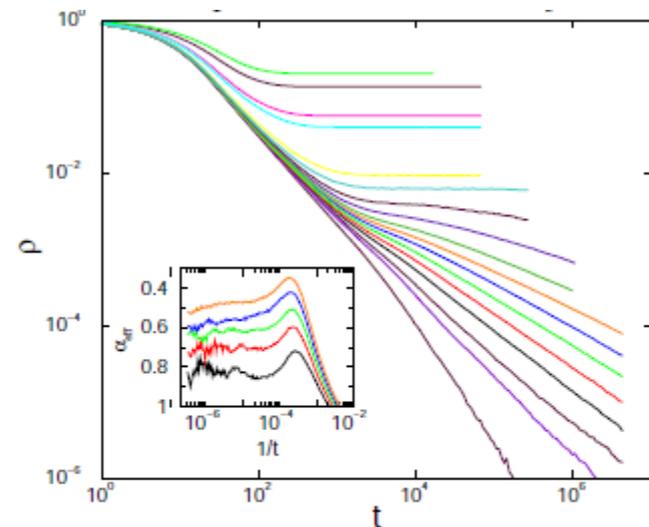


FIG. 5: Density decay in model B $m = 1$, $m_0 = 20$, $N = 10^5$ for $\lambda = 6.8, \dots, 15$ (top to bottom). Inset: corresponding local slopes in the GP region.

Heterogeneous mean-field theory: critical point, with linear density decay: $\rho \propto 1/t$
can't describe frozen disorder !

Do power-laws survive the thermodynamic limit ?

- Finite size analysis shows the disappearance of a power-law scaling:

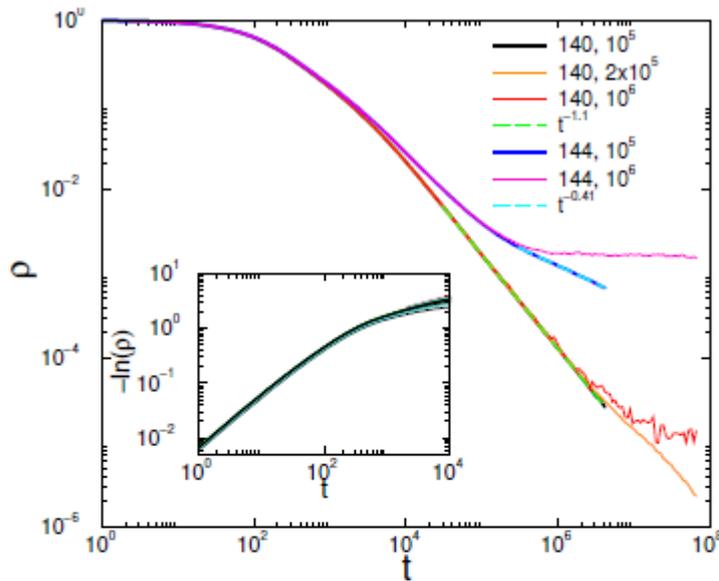


FIG. 5. Density decay as a function of time $\rho(t)$ for the CP on weighted BA trees with a multiplicative weighting scheme (WBAT-I) with exponent $\nu = 1.5$. Plots correspond to two sets of λ (upper branch: $\lambda = 144$, lower branch $\lambda = 140$) at different network sizes N . Dashed lines represent PL fittings. Inset: Initial time region of the same data, showing an stretched exponential behavior.

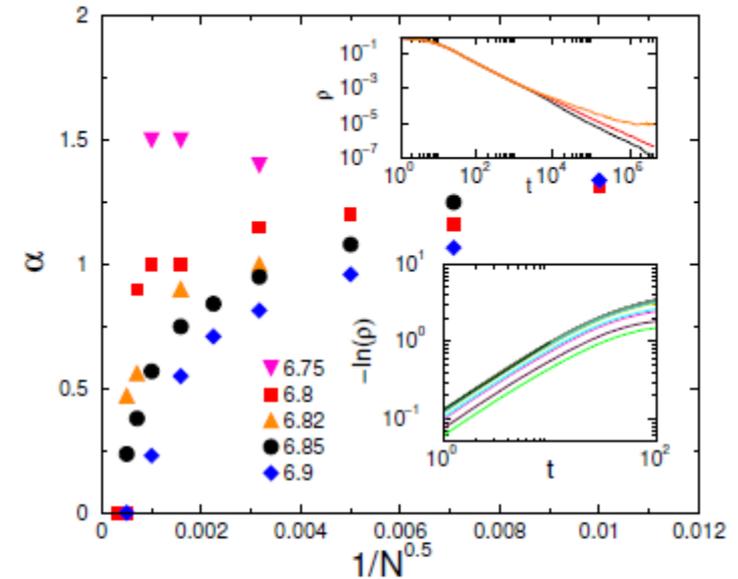


FIG. 8. Finite-size scaling analysis of the density decay exponent for $\lambda = 6.75$ (triangles), $\lambda = 6.8$ (boxes), $\lambda = 6.82$ (triangles), $\lambda = 6.85$ (bullets), $\lambda = 6.9$ (rhombes) in the CP on weighted BA trees with an age-dependent weighting scheme (WBAT-II) with exponent $x = 2$. Top inset: $\rho(t)$ for $\lambda = 6.82$ ($N = 10^6$, $N = 4 \times 10^5$, $N = 10^5$ top to bottom). Bottom inset: Initial time density.

- Smearred phase transition: power-law \rightarrow saturation:
- **Rare sub-spaces, but infinite dimensional ?**

Percolation analysis of the weighted BA tree

We consider a network of a given size N , and delete all the edges with a weight smaller than a threshold ω_{th} .

For small values of ω_{th} , many edges remain in the system, and they form a connected network with a single cluster encompassing almost all the vertices in the network.

When increasing the value of ω_{th} , the network breaks down into smaller subnetworks of connected edges, joined by weights larger than ω_{th} .

The size of the largest ones grows linearly with the network size N

↔ standard percolation transition.

These clusters, which can become arbitrarily large in the thermodynamic limit, play the role of correlated **RRs**, sustaining independently activity and smearing down the phase transition.

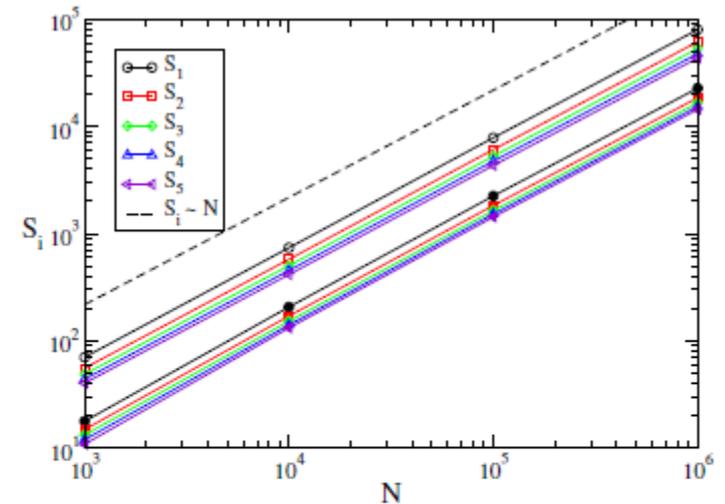


FIG. 6. Size S_i of the 5 largest clusters in a percolation analysis of the WBAT-I model with $\nu = 1.5$ for $\omega_{th} = 100\omega_{min}$ (hollow symbols) and $\omega_{th} = 1000\omega_{min}$ (full symbols), where ω_{min} is the minimum weight in the network. The size of all components grows linearly with network size N , and is therefore infinite in the thermodynamic limit.

Summary

- Quenched disorder in complex networks can cause slow dynamics :
Rare-regions \rightarrow (Griffiths) phases \rightarrow *no tuning or self-organization needed !*
- In **finite dim.** (for CP) GP can occur *due to topological disorder*
- In **infinite dim**, scale-free, BA network mean-field transition of CP with logarithmic corrections (HMF+simulations)
- In BA **trees** non mean-field transition observed
- In **weighted BA trees** non-universal, slow, power-law dynamics can occur for finite N , but in the $N \rightarrow \infty$ limit saturation is observed
- Smearred transition can describe this, percolation analysis confirms the existence of arbitrarily large dimensional sub-spaces with (correlated) large weights
- Acknowledgements to : HPC-Europa2, OTKA, Osiris FP7

[1] M. A. Munoz, R. Juhasz, C. Castellano, and G. Ódor, *Phys. Rev. Lett.* 105, 128701 (2010)

[2] G. Ódor, R. Juhasz, C. Castellano, M. A. Munoz, *AIP Conf. Proc.* 1332, Melville, New York (2011) p. 172-178.

[3] R. Juhasz, G. Ódor, C. Castellano, M. A. Munoz, *Phys. Rev. E* 85, 066125 (2012)

[4] G. Ódor and Romualdo Pastor-Satorras, *Phys. Rev. E* 86, 026117 (2012)