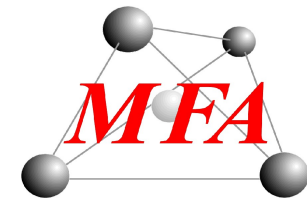


Griffiths phases of contact processes in complex networks

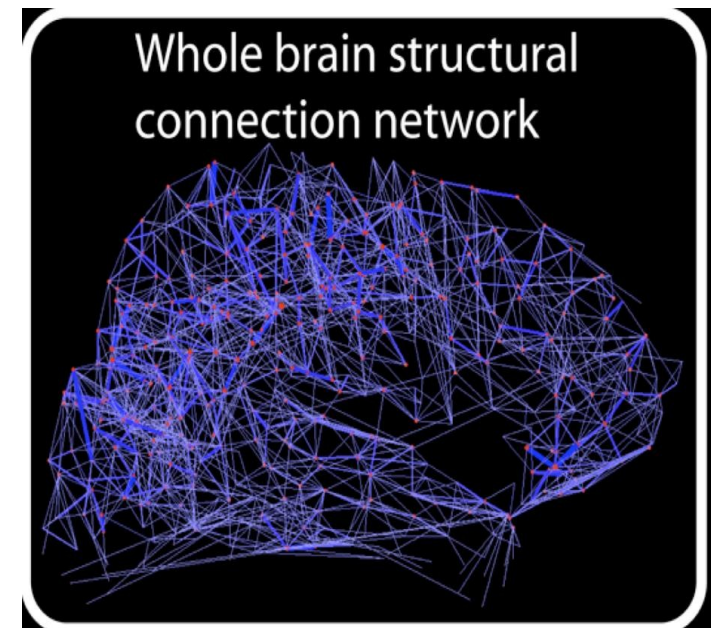
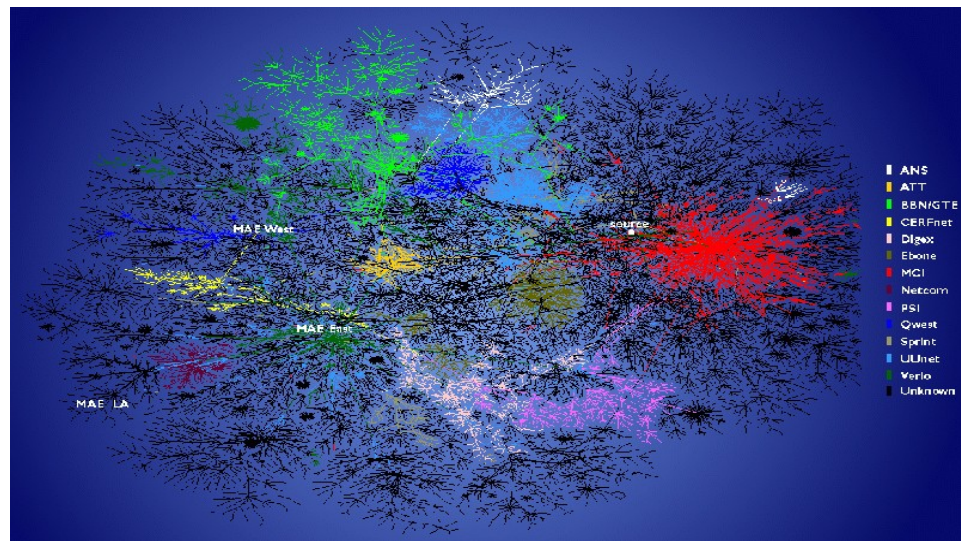
Géza Ódor

RESEARCH INSTITUTE FOR TECHNICAL PHYSICS AND MATERIALS SCIENCE (MFA) BUDAPEST



- Exploration of complex networks is flourishing since ~2000 (Barabási & Albert)
- Dynamical systems living on networks is of current interest
- Research of disordered materials is intensifying
- Open: Complex networks + quenched disorder ?
- Origin of non-universal (dynamic) scaling behaviors ?

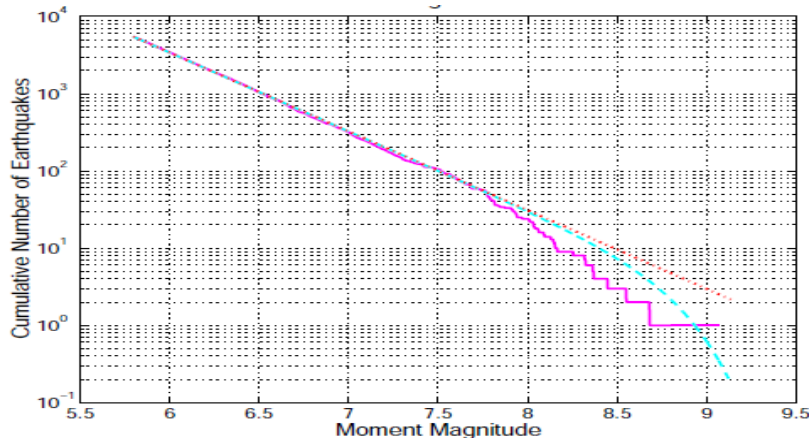
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Diffusion spectrum imaging

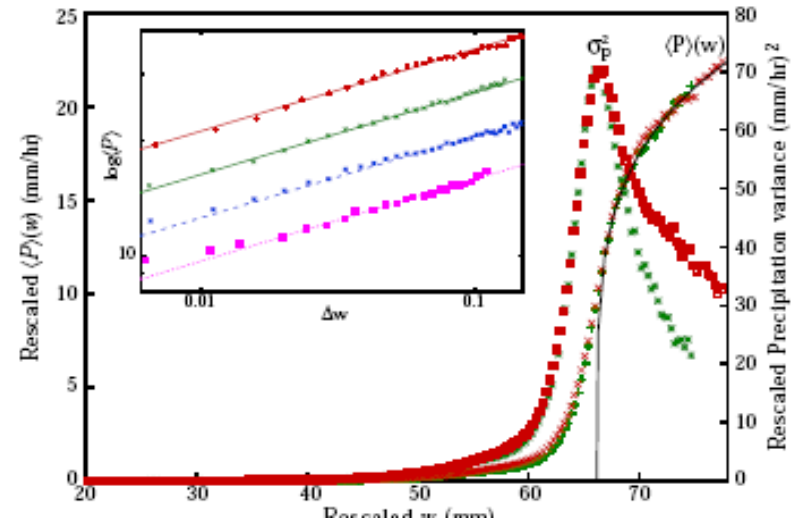
Power-law scaling in nature

- Earthquake size distribution



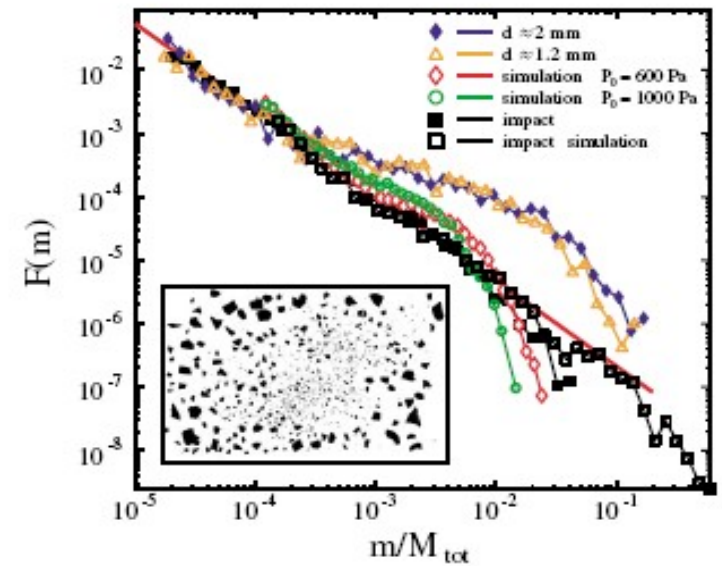
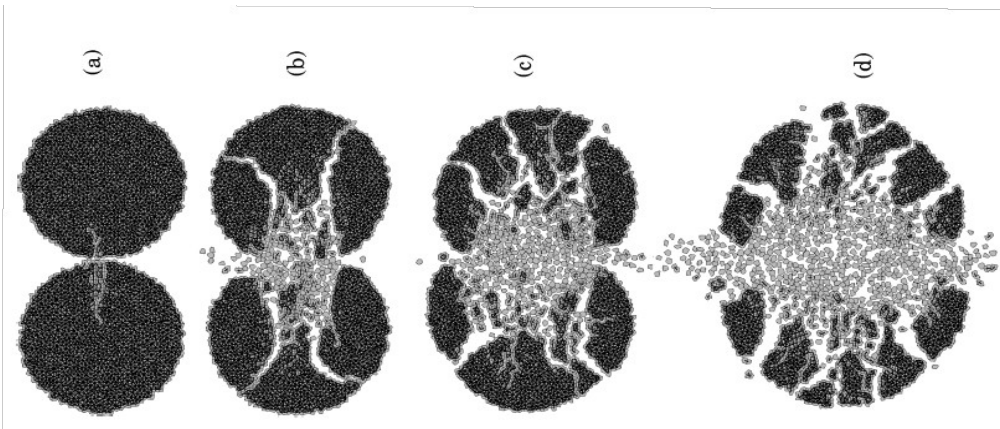
Meteorology and Climatology:

O. Petres and D. Neelin, Nature Phys. 2 (2006) 393



Damage formation by collisions, explosions

(F. Kun & H. Hermann)



- And many more...

What is the origin ?

“Self-Organized Criticality” (SOC) ?

Bak-Tang-Wiesenfeld sand pile model (1987) CA :

Add a grain of sand:

$$h(x,y) \rightarrow h(x,y) + 1$$

And avalanche if: $h(x,y) > h_c$:

$$h(x, y) \rightarrow h(x, y) - 4$$

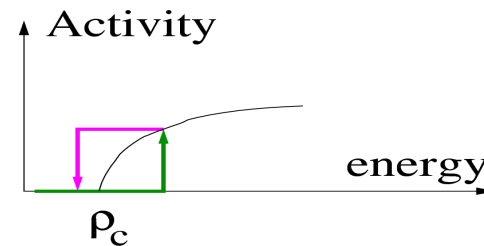
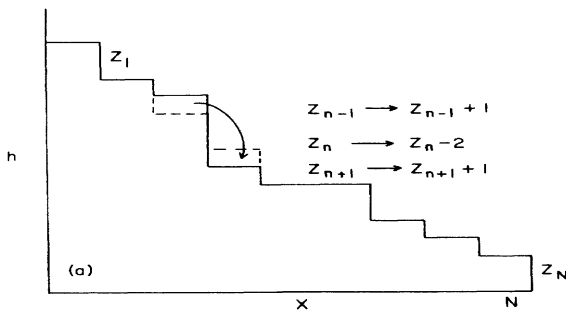
$$h(x \pm 1, y) \rightarrow h(x \pm 1, y) + 1$$

$$h(x, y \pm 1) \rightarrow h(x, y \pm 1) + 1$$

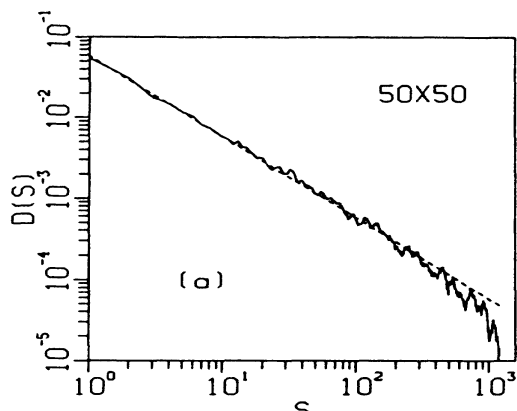
The term SOC usually refers to a mechanism of *slow energy accumulation* and *fast energy redistribution*, driving a system toward a critical state.

← **Prototype: sand pile model**

Self-tuning to critical point:



Stochastic sand pile models can be mapped onto ordinary **nonequilibrium criticality** at **phase transition to absorbing phases** (*Dickman et al. 2000*)

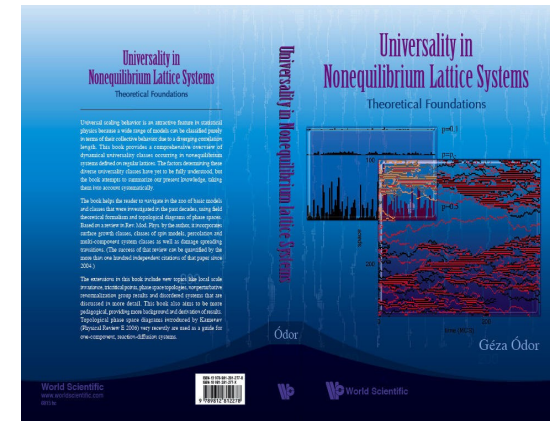


Scaling-universality classes of basic models

Universality classes in complex system due to : $\xi \rightarrow \infty$
 i.e: near **critical points** or in **system with currents (nonequilibrium)**

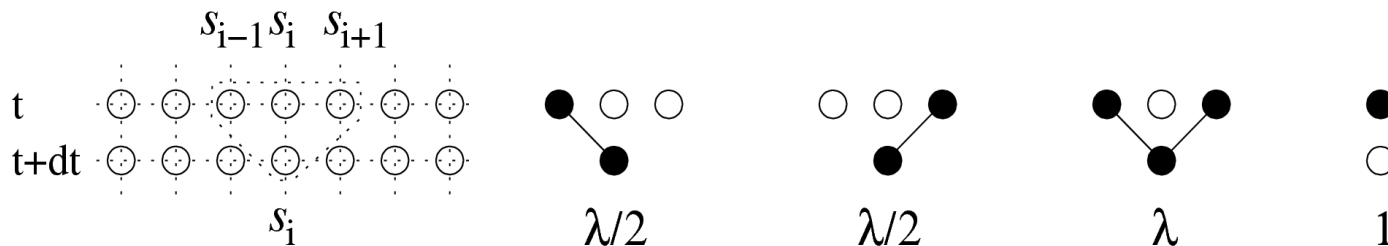
Basic models can be classified by universal scaling behavior according to some global conditions: dimensions, symmetries, ..., topological constraints :

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



What prevents the observation of universal behavior ?

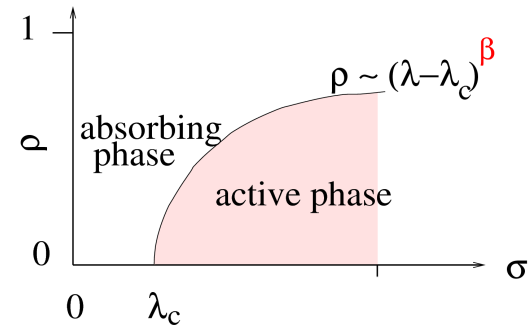
Prototype model describing epidemic or activity spreading : **Contact Process (CP) (1d)**:



Contact process, scaling, DP

- **Order parameter:** density of active sites: $\rho(t) = \frac{1}{L^d} \sum_{\mathbf{r}} \langle n_{\mathbf{r}}(t) \rangle$ $\rho_{\text{stat}} = \lim_{t \rightarrow \infty} \rho(t)$

- **Critical phase transition**
between active \leftrightarrow absorbing states:



- **Scaling behavior**

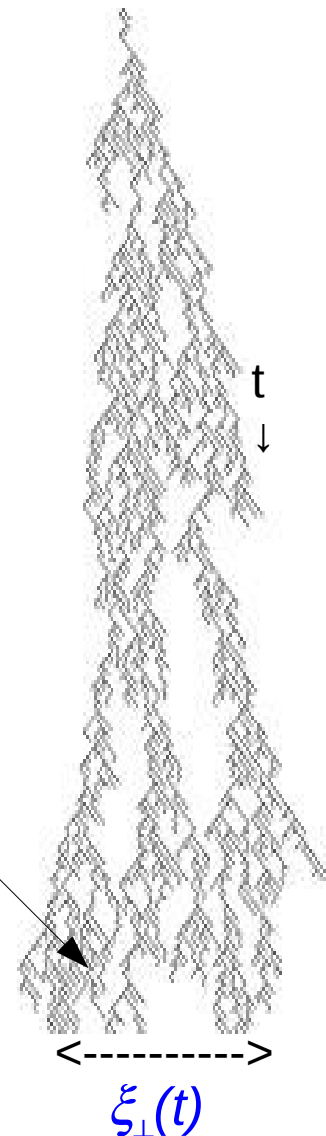
Correlation length diverges anisotropically : $\xi_{\perp} \sim |\Delta|^{-\nu_{\perp}}$, $\xi_{\parallel} \propto |\Delta|^{-\nu_{\parallel}}$

Independent critical exponents: β , ν_{\perp} , ν_{\parallel}

Density decay: $\rho(t) \sim t^{-\alpha}$ where: $\alpha = \beta / \nu_{\parallel}$

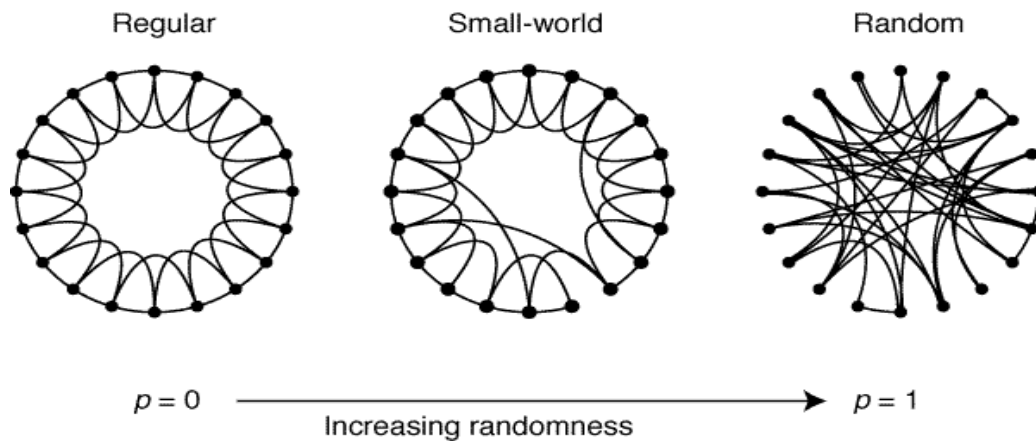
- **Directed Percolation, DP** Universality Class

Robust class of epidemic, information, ... etc. spreading
well known in $d = 1, 2, 3, 4$... Euclidean dimensions



Basic network models

From regular to random networks:



Erdős-Rényi ($p = 1$)

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

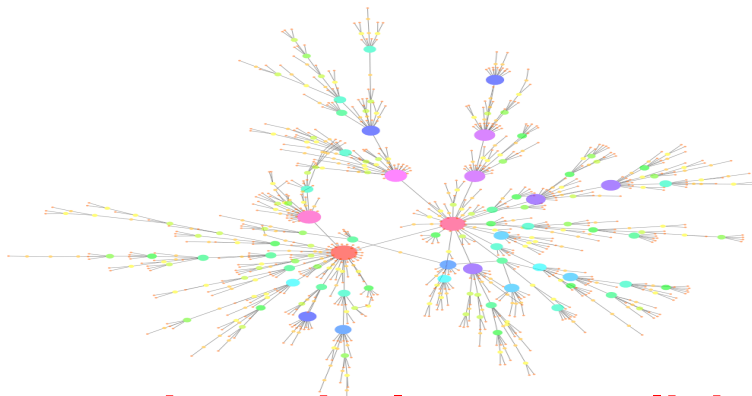
$$P(k) = e^{-\langle k \rangle} \langle k \rangle^k / k!$$

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

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

Below: $d = 0$

Scale free networks:



Barabási-Albert

Degree distribution:

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

Topological dimension: $d = \infty$

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

How do networks differ from Euclidean lattices in general ?

1) **Small diameter** → long-range interactions → weak fluctuations → mean-field behavior

DP + Lévy flight : $P(r) \sim r^{-(d+\sigma)}$

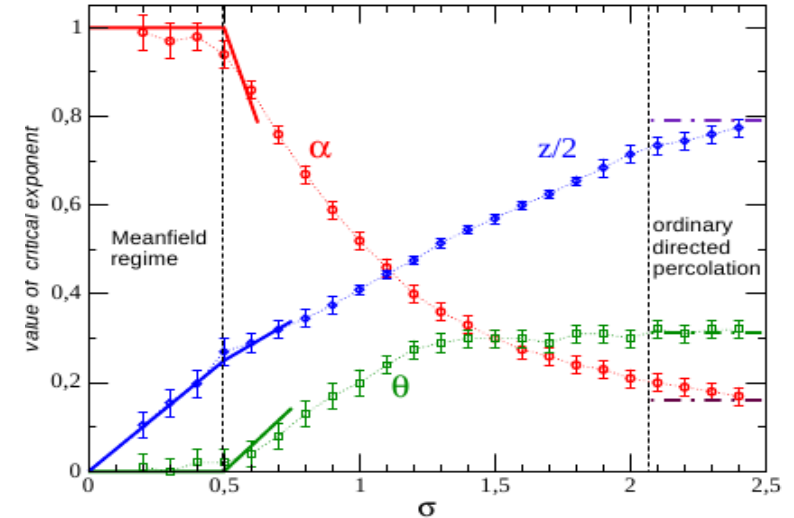
Continuously changing exponents :

DP → mean-field-**DP** as: $\sigma \rightarrow 0$



2) **Heterogenicity** (i.e. aperiodicity, changing degree k ...) → Other critical points ...

3) **Disorder (quenched)** → Other critical points ...
Griffiths phases, activated scaling ?



(Janssen et al, Hinrichsen et al 1999)

Up to now mainly **1)** + **2)** cases have been investigated by dynamical models

The heterogeneous mean-field approximation omits **quenched disorder**,
describes annealed disorder, what we have learnt :

Topology is relevant :



The effect of topological disorder on scale-free: $P(k) \sim k^{-\gamma}$ networks:

$\lambda_c = \langle k \rangle / \langle k^2 \rangle = 0$ if $2 < \gamma < 3$ (Ising, CP ...)

Always active Contact Process;

$\lambda_c > 0$ if $\gamma = 3$, or finite size, or weakened (weighted) links:

γ dependent exponents

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

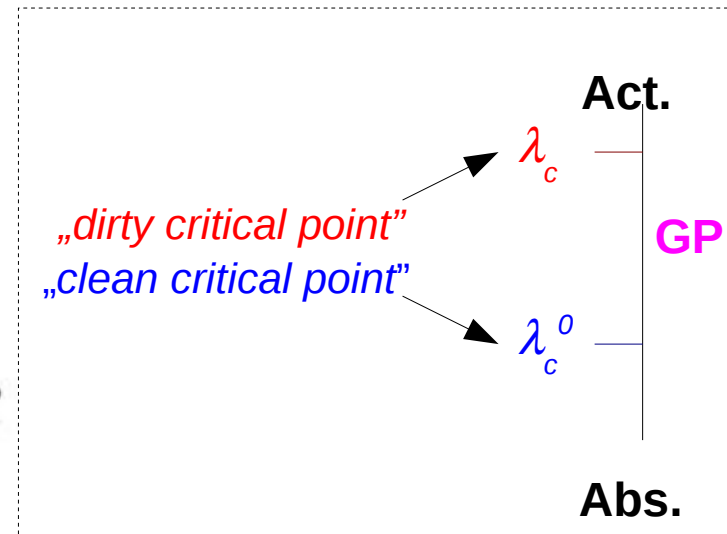
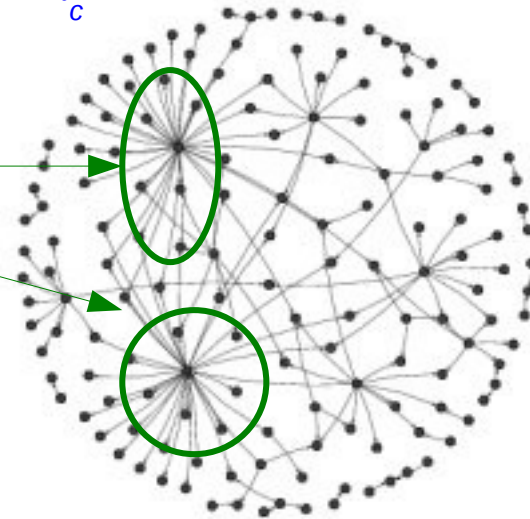
- At λ_c^0 the characteristic time scales as: $\tau(L_R) \sim L_R^{-z} \Rightarrow$ saddle point analysis:

$$\ln \rho(t) \sim t^{d / (d+z)} \quad (\text{stretched exponential})$$

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

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

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



Griffiths Phase

(continuously changing exponents)

CP + Topological disorder results

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

• Top. dim: $N(l) \sim l^d$ $d(\beta)$ **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** %

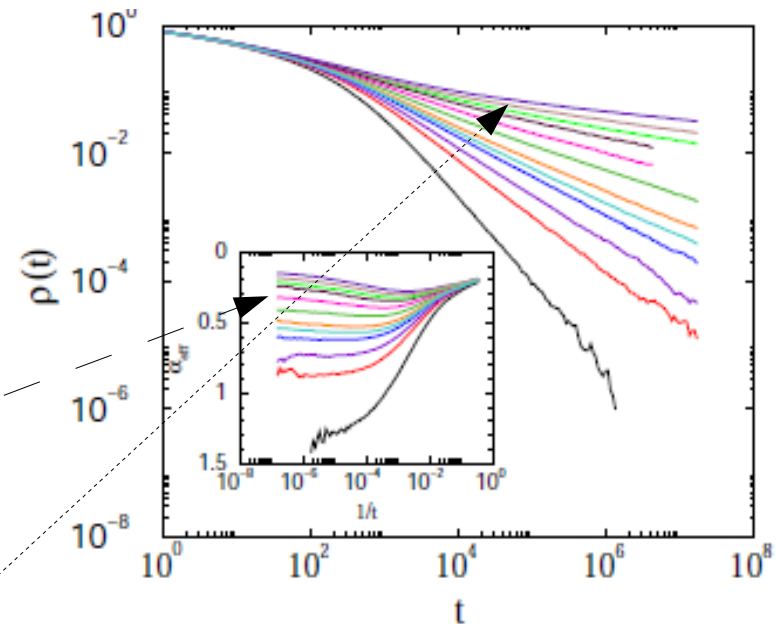
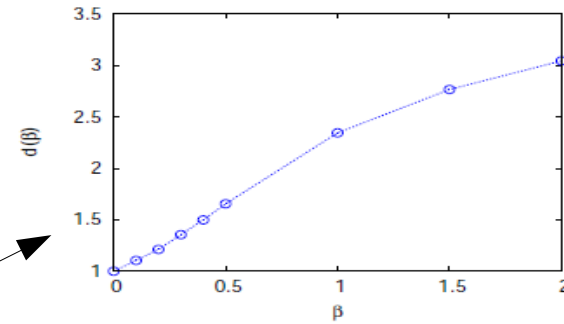
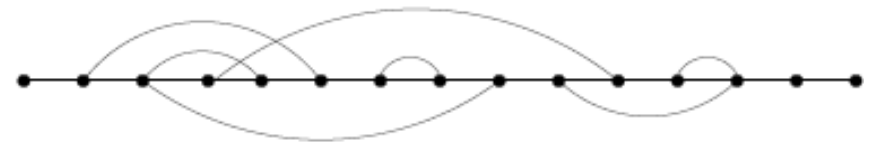


FIG. 3: Density decay in Benjamini-Berger networks with $s = 2$ and $\beta = 0.2$ for different values of λ (from top to bottom: 2.81, 2.795, 2.782, 2.77, 2.75, 2.73, 2.71, 2.70, 2.69, 2.67, 2.65, 2.6). Straight lines lie in the Griffiths phase. Inset: Corresponding effective exponents, illustrating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

Generalized SW networks with GP

- CP on **cubic regular random networks**
 $L = 10^7$ sized simulations:
Griffiths Phase (strong log. corrections)

- For *general*: $P(l) \sim \beta l^{-s}$

$s = 0$ mean-field behavior

$s < 2$ $d = \infty$, power-laws (Lévy) with no GP

$s > 2$ mean edge length is finite,

$d = 1$ CP with topological disorder
rare regions: subgraphs with more internal edges than the average

Griffiths Phase

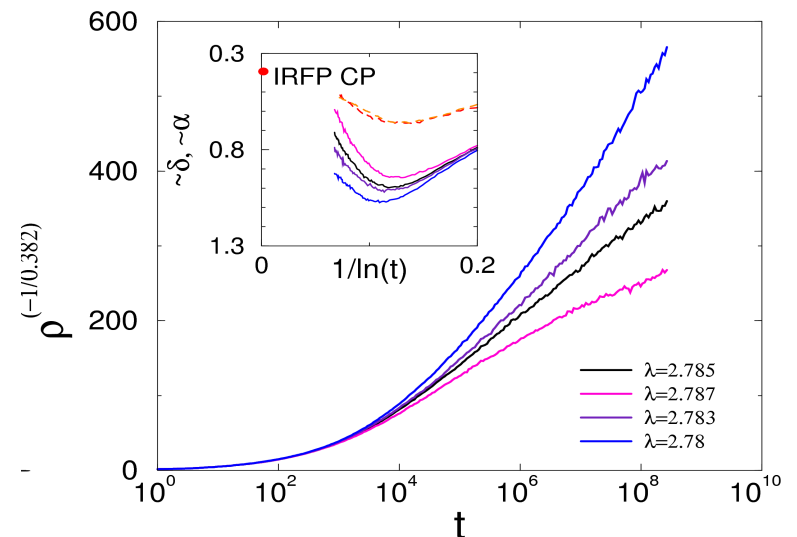
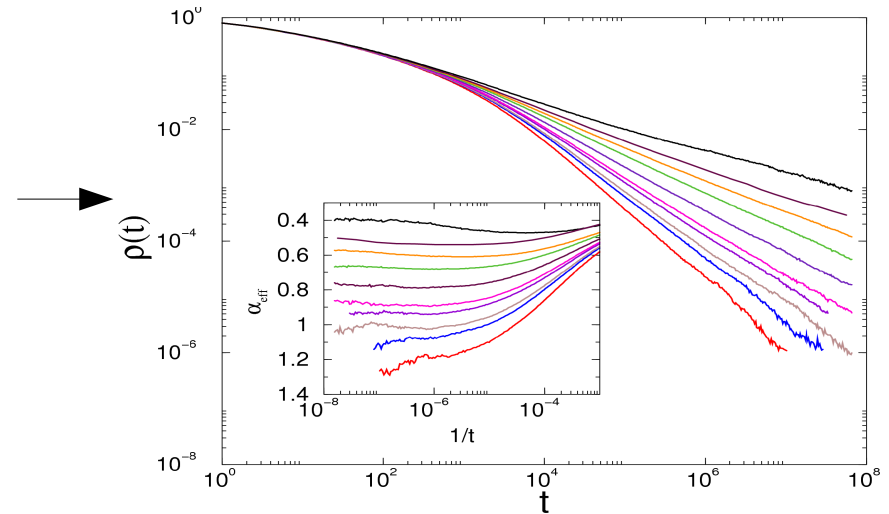
Conjecture: IRPF of the disordered CP

$$\rho(t) \sim [\ln t]^{-\bar{\delta}} \quad \bar{\delta} = 0.38197\dots$$

$$[\rho(t)]^{-1/\bar{\delta}} \sim \ln t$$

$$\ln \rho(t) \sim -\bar{\delta} \ln(\ln t)$$

$$\bar{\delta}_{\text{eff}}(t) = -\frac{d \ln \rho}{d \ln(\ln t)}$$



For other networks ?

Above percolation threshold rare regions do not emerge →

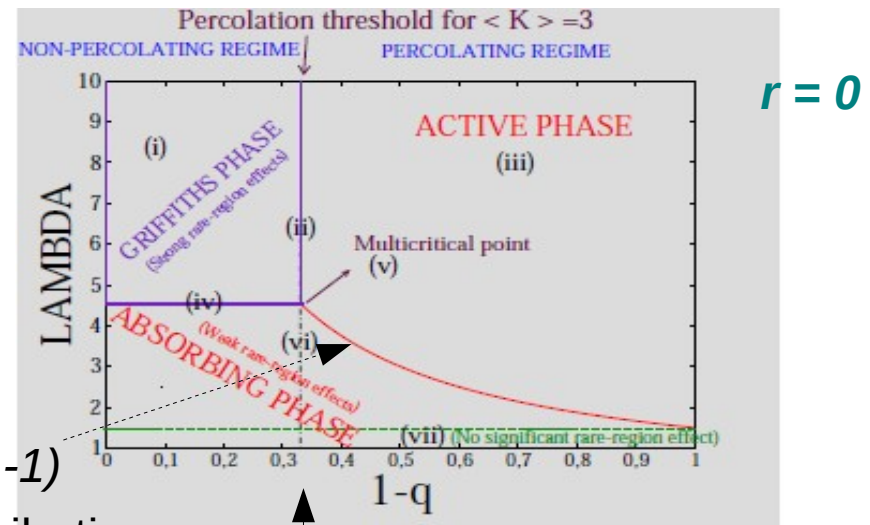
Do not expect Griffiths phase in nets like Barabási-Albert vagy Erdős-Rényi if $\langle k \rangle$ is large

Q-CP on Erdős-Rényi network $\langle k \rangle = 3$

Mean-field : $\lambda_c = 1$, Pair mean-field: $\lambda_c^2 = \langle k \rangle / (\langle k \rangle - 1)$

+ reduced reaction rates with binary probability distribution:

$$P[\lambda(x)] = q \delta[\lambda(x) - r\lambda] + (1 - q) \delta[\lambda(x) - \lambda]$$



$r = 0$

$$q_{perc} = 2/3$$

Rare Region theory

(i) GP. Cluster size distribution : $P(s) \sim s^{-3/2} \exp[-s(p-1-\ln(p))]$, where: $p = \langle k \rangle (1-q)$

$$\rho(t) \sim \int ds s P(s) \exp[-t/\tau(s)]$$

$$\text{where: } \tau(s) \cong t_0 \exp[A(\lambda)s]$$

$$\lambda_c^2 = \langle k \rangle / (\langle k \rangle - 1) [1 - q(1-r)]$$

$r = 0$ case:

Saddle-point approximation: $\rho(t) \sim t^{-\gamma}$ $\gamma = (p-1-\ln(p)) A(\lambda)$ cont. changing dyn. exps.

(ii) $q = q_{perc}$: $\rho(t) \sim [\ln(t/t_0)]^{-1/2}$ ha $\lambda > \lambda_c$ activated scaling

(iii) Active phase: Giant component, MF+Griffiths eff., If $q < q_{perc}$ critical line: $\rho(t) \sim 1/t$

(iv) $q > q_{perc}$ $\lambda = \lambda_c \exists$ RR almost critical, stretched exponential decay

(v) Multi-critical point: power-law decay

(vi) Abs phase: $\lambda_c(q_{perc}) > \lambda > \lambda_c(q=0) \exists$ supercritical RR if: $q_{loc} < q_{perc}$, but weak effects

(vii) $\lambda < 1.5$: RR free region : exponential decay

$r > 0$: Similar phase-diagram, but active phase goes up to $q=1$

Simulation results for Q-CP on Erdős-Rényi networks

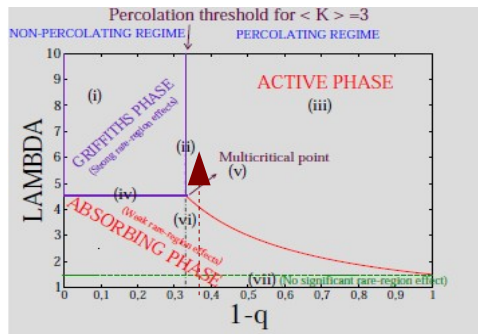
$N \leq 10^7$ $\langle k \rangle = 3$ Decay from fully active state,
 Cluster growth from a single seed
 Same asymptotic behavior : *Rapidity* symmetry

Clear case: $\lambda_c(q=0) \approx 1.5$, $\lambda_c(q=1) \approx 30$ agrees with:

$$\lambda_c^2 = \langle k \rangle / (\langle k \rangle - 1) [1 - q(1-r)] \quad \text{pair approximation,}$$

$$\rho(t) \sim 1/t$$

$0 < q < 1$: Agreement with the theory:



$r = 0$, $q < 2/3$: Mean-field DP critical transition

$q > 2/3$: No active phase:

$$\lambda > \lambda_c(q = q_{perc} = 2/3) = 4.5: \quad \rho(t) \sim t^{-\gamma} \quad \text{GP}$$

$r > 0$: similar phase diagram,
 active phase goes up to $q = 1$

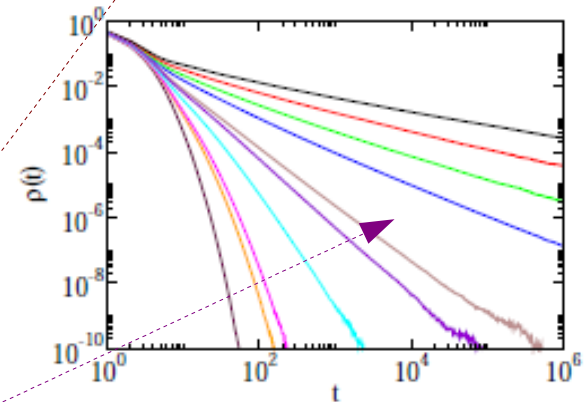
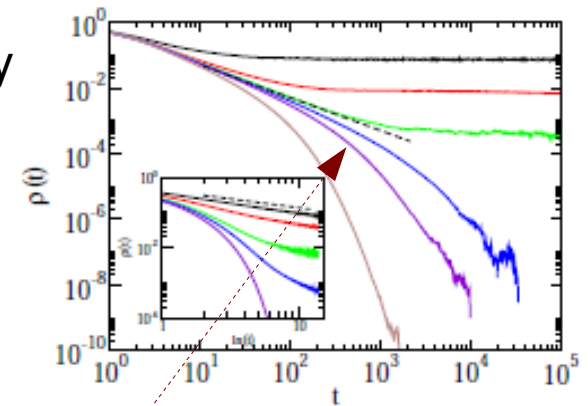


FIG. 1: Average activity density $\rho(t)$ vs time t for ER networks with $\langle k \rangle = 3$, $r = 0$, and $N = 10^5$. λ s are ordered from top to bottom in all panels. (a) Upper panel: $q = 0.6$, and $\lambda = 5, 3.8, 3.6, 3.55, 3.5, 3.3$. The dashed line is proportional to t^{-1} . (a) Inset: ρ vs $\ln(t)$ for $q = 2/3$; $\lambda = 10, 7, 5, 4.5, 4$; the dashed line is proportional to $\ln(t)^{-1/2}$. (b) Lower panel: $q = 0.9$, and $\lambda = 50, 30, 20, 15, 10, 9, 7, 5, 4.5, 2.7$. Straight lines lie in the Griffiths phase.

Summary

- Quenched **disorder** increases the occurrences of **non-universal scaling behavior**: **Extended Griffiths Phases must occur in nature**
- **Slow** (algebraic, logarithmic) dependence is expected in many dynamical systems defined on networks (\leftrightarrow **Fast** dynamics in **pure** network models)

Understanding algebraic slow forgetting times of working memory in a **simple model of brain** (see the paper/talk by Johnson et al)

In other epidemic/information spreading models living on heterogeneous random networks with **finite topological dimensions**,

- New Hungarian-Spanish-Italian collaboration has been set up with the first results:

Griffiths phases on complex networks: M. A. Munoz, R. Juhász, C. Castellano and G. Ódor
arXiv:1009.0395, PRL in press.

- Acknowledgments: **HPC2 Europe, OTKA, OSIRIS FP7**

