

TREC Evaluation 2008-2011

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INRIA & ENS

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DISCRETE GEOMETRY: RANDOM GRAPHS

- Mathematical tools:

- Applications:

DISCRETE GEOMETRY: RANDOM GRAPHS

- Mathematical tools:
 - Cavity method
 - Configuration model
- Applications:

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- Mathematical tools:
 - Cavity method
 - Configuration model
- Applications:
 - Pure Mathematics:
 - * Random matrices
 - * Combinatorial optimization
 - * First passage percolation
 - Communication Networks:
 - * Load balancing
 - Network economics:
 - * Diffusions in social networks
 - * Control of epidemics
 - * Mean field games for security

DISCRETE GEOMETRY: RANDOM GRAPHS

- Mathematical tools:
 - Cavity method
 - Configuration model
- Applications:
 - Pure Mathematics:
 - * Random matrices (SODA, RSA, AoP)
 - * Combinatorial optimization (AIHP, SIAMJComp, PTRF)
 - * First passage percolation
 - Communication Networks:
 - * Load balancing (SODA, SIGMETRICS)
 - Network economics:
 - * Diffusions in social networks (GEB)
 - * Control of epidemics (SIGMETRICS)
 - * Mean field games for security (INFOCOM(3), SIGMETRICS)

DISCRETE GEOMETRY: RANDOM GRAPHS

- Mathematical tools:
 - Cavity method
 - Configuration model
- Applications:
 - Pure Mathematics:
 - * Random matrices C. Bordenave, J. Salez
 - * Combinatorial optimization D. Aldous, C. Bordenave, J. Salez
 - * First passage percolation M. Draief, H. Amini
 - Communication Networks:
 - * Load balancing M. Leconte, L. Massoulié
 - Network economics:
 - * Diffusions in social networks E. Coupechoux
 - * Control of epidemics
 - * Mean field games for security J. Bolot

WEIGHTED DIAMETER

Graph $G = (V, E)$:

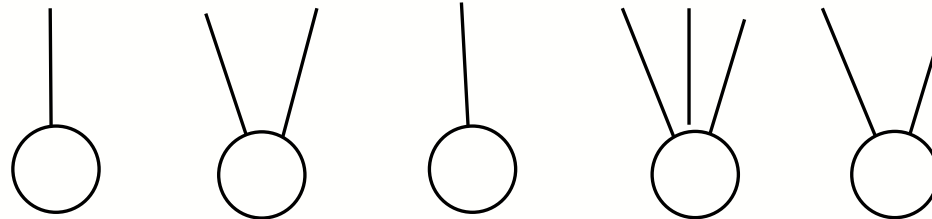
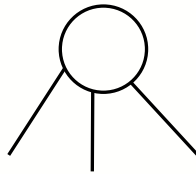
- Distance $\text{dist}(a, b) = \min_{\pi \in \Pi(a, b)} |\pi|$, the number of edges in E in the shortest path connecting a and b .
- Diameter of G defined by:

$$\text{diam}(G) = \max\{\text{dist}(a, b), a, b \in V, \text{dist}(a, b) < \infty\}.$$

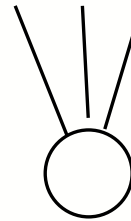
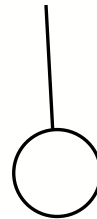
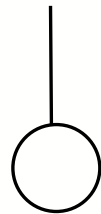
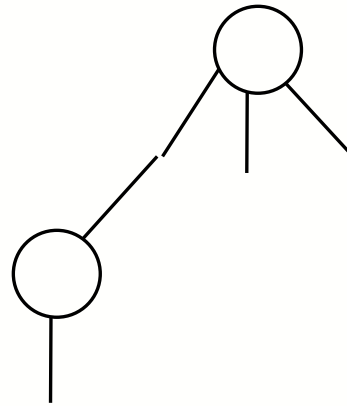
- Weight associated to each edge $e \in E$: w_e .
- Weighted distance $\text{dist}_w(a, b) = \min_{\pi \in \Pi(a, b)} \sum_{e \in \pi} w_e$.
- Weighted diameter of G defined by:

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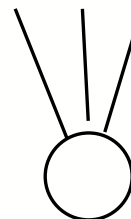
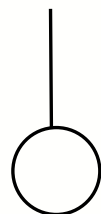
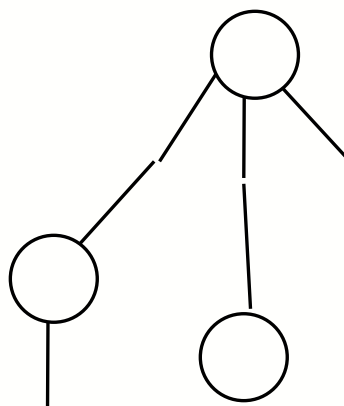
EXPLORATION PROCESS



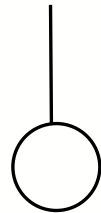
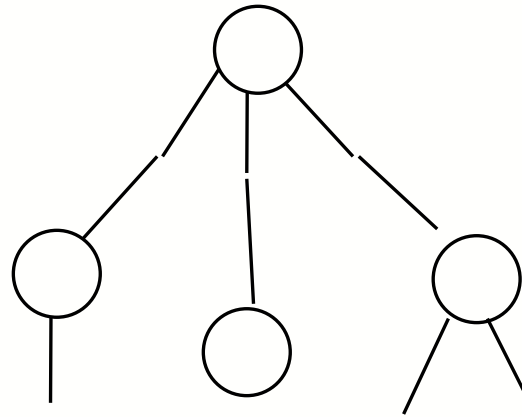
EXPLORATION PROCESS



EXPLORATION PROCESS



EXPLORATION PROCESS



BRANCHING PROCESS APPROXIMATION

The first individual has offspring distribution $\{p_k\}$.

The other individuals have offspring distribution $\{q_k\}$.

Let $\{q_k\}_{k=0}^{\infty}$ the size-biased probability mass function corresponding to $\{p_k\}$, by

$$q_k = \frac{(k+1)p_{k+1}}{\lambda}, \text{ and, } \nu = \sum_{k=0}^{\infty} kq_k \in (0, \infty).$$

The mean of the size of generation k is $\lambda\nu^{k-1}$.

The condition $\nu > 1$ is equivalent to the existence of a giant component.

TYPICAL GRAPH DISTANCE

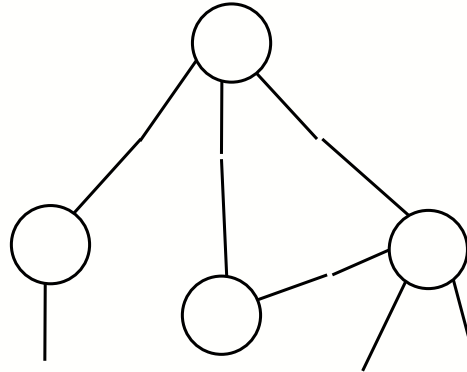
Theorem 1. *For a and b chosen uniformly at random in the giant component of $G(n, (d_i)_1^n)$, we have*

$$\frac{\text{dist}(a, b)}{\log n} \xrightarrow{p} \frac{1}{\log \nu}.$$

Van der Hofstad, Hooghiemstra, Van Mieghem 2005

A SIMPLE HEURISTIC

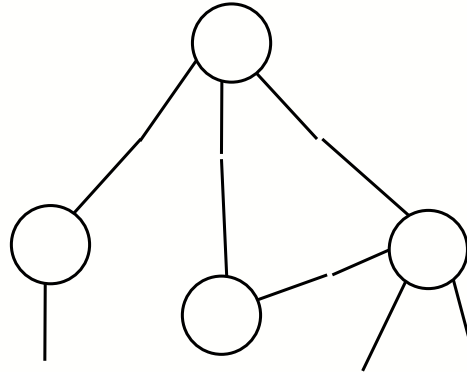
Let $Z_k^{(1)}$ be the number of free half-edges in the ball $B(a, k) = \{i, \text{dist}(1, i) \leq k\}$.
 $Z_0^{(1)}$ is the degree of node 1.



By the branching process approximation, $Z_k^{(1)}$ is close to $\lambda \nu^{k-1}$.

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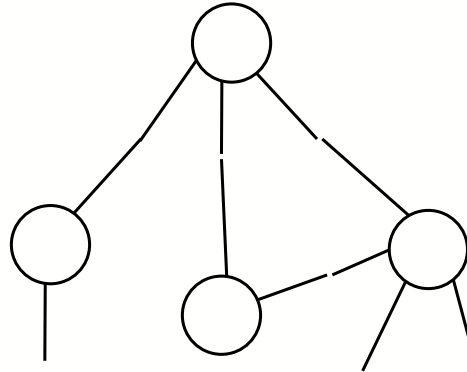


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A free half-edge of $Z_k^{(1)}$ is attached to a free half-edge of $Z_k^{(2)}$ with positive probability if $Z_k^{(1)} Z_k^{(2)}$ is of order the total number of free half-edges left after k exploration steps.

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Take $k \approx \frac{1}{2} \frac{\log n}{\log \nu}$, then $Z_k^{(1)} \approx Z_k^{(2)} \approx \sqrt{n}$ and the number of free half-edges is $\approx n - 2\sqrt{n} \approx n$.

The typical distance between 1 and 2 is $\approx 2k = \frac{\log n}{\log \nu}$.

WEIGHTED DIAMETER

Theorem 2. Consider a random graph $G(n, (d_i)_1^n)$ with i.i.d. exponential 1 weights on its edges, then

$$\frac{\text{diam}_w(G(n, (d_i)_1^n))}{\log n} \xrightarrow{p} \frac{1}{\nu - 1} + \frac{2}{d_{\min}} \mathbf{1}_{(d_{\min} \geq 3)} + \frac{\mathbf{1}_{(d_{\min}=2)}}{1 - q_1} + \frac{2}{1 - \beta_*} \mathbf{1}_{(d_{\min}=1)}.$$

Ding, Han Kim, Lubetzky, Peres 2010 (random regular graphs)

Amini, Draief, L. 2011

Key ideas:

- Coupling with a continuous time branching process
- Large deviations for split times

WEIGHTED DIAMETER

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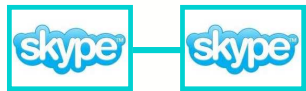

Amini, Draief, L. 2011

Key ideas:

- Coupling with a continuous time branching process
- Large deviations for split times

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{\log n} \log \mathbb{P} \left(\infty > T_n^k \geq \left(x + \frac{1}{\nu - 1} \right) \log n \right) &= -x g(\xi_{\min}, k) \\ \mathbb{P} \left(T_n^k < \left(\frac{1 - x}{\nu - 1} \right) \log n \right) &= o \left(n^C e^{-n^x} \right). \end{aligned}$$

Game-theoretic contagion model

Situation	Payoff (for both users)
	q
	$1 - q > q$
	0

Game on a network of interconnected players **Blume 1995, Morris 2000**

- Total payoff = sum of payoffs from all your neighbors

- Switch from  to  $\Leftrightarrow \frac{|\text{Neighbors using Skype}|}{|\text{Neighbors}|} > q$.

Contagion in networks

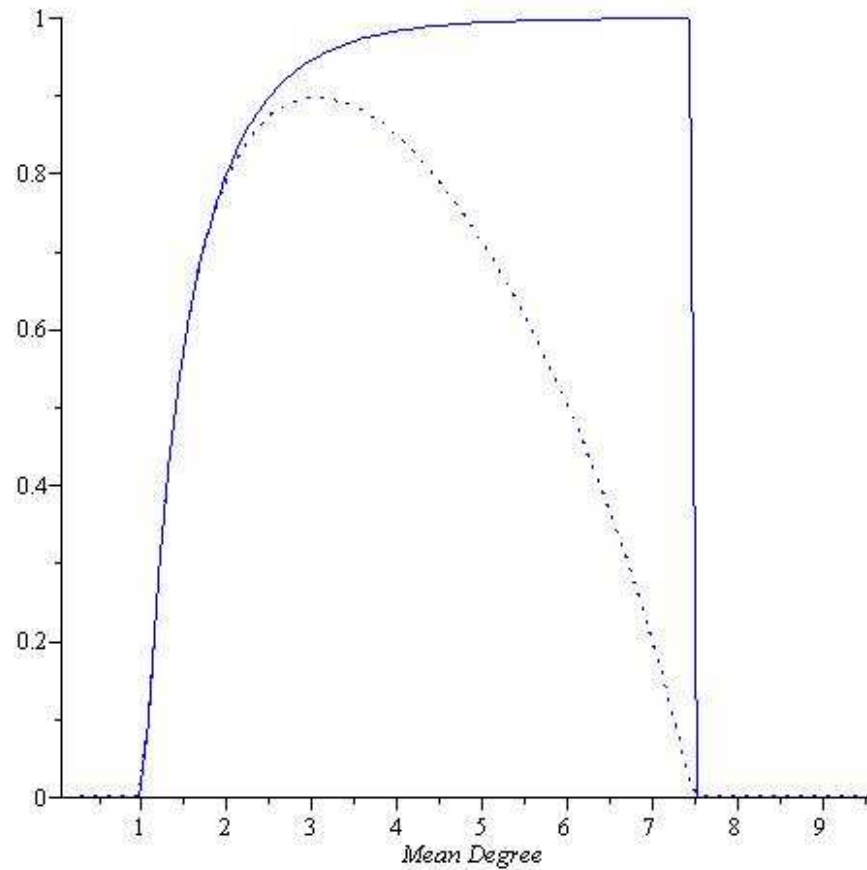
Parameter q varies:

q small \Rightarrow CASCADE

q higher \Rightarrow NO cascade

Contagion threshold $q_c^{(G)} := \sup \{ q \mid \text{CASCADE in } G \text{ for parameter } q \}$

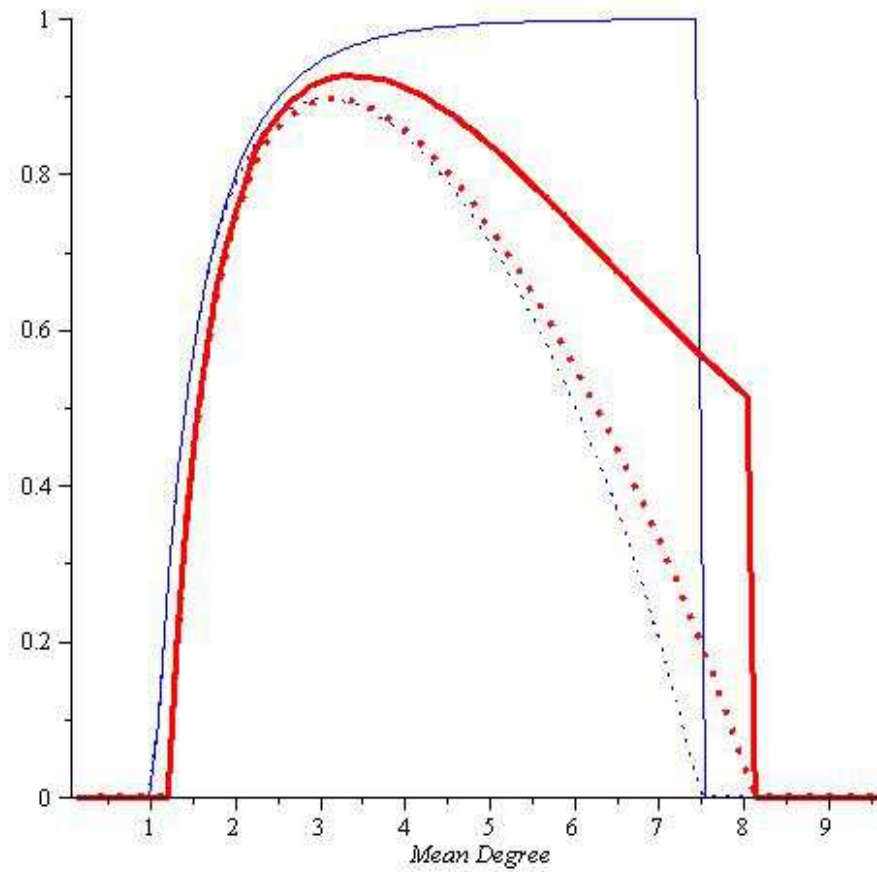
Contagion in random networks: impact of connectivity



... Pivotal players in the random graph

— Cascade size in the random graph

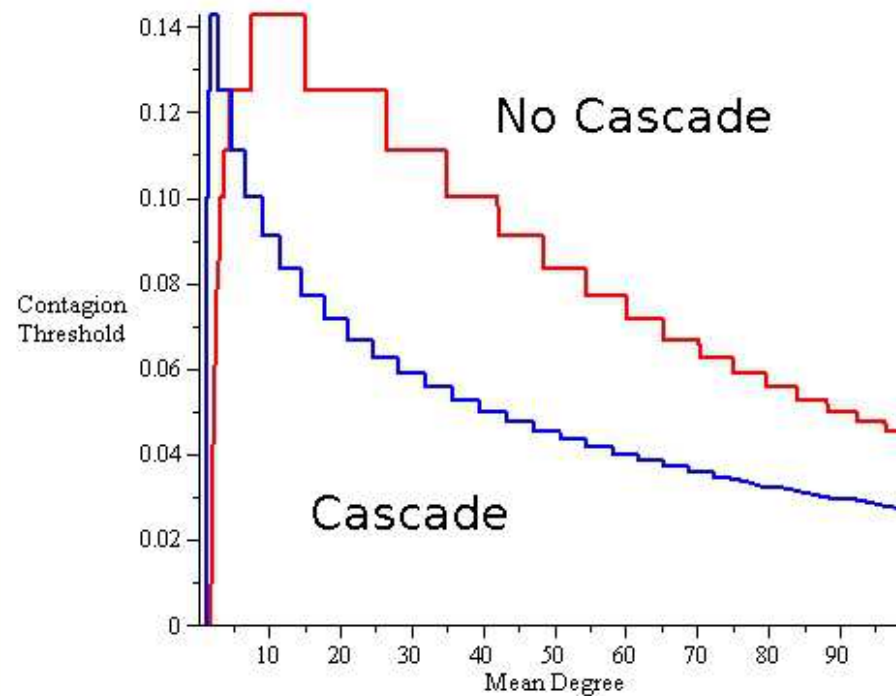
Contagion in random networks with clustering



joint work with E. Coupechoux

Contagion in random networks with clustering

Graphs with the SAME asymptotic degree distribution: $\tilde{p}_k \propto k^{-\tau} e^{-k/50}$



- Graph with clustering
- Graph with no clustering

joint work with E. Coupechoux

FUTURE

- Mathematical tools:
 - Cavity method
 - Configuration model
- Applications:
 - Pure Mathematics:
 - * Random matrices
 - * Combinatorial optimization
 - * First passage percolation → stochastic order, Bartek
 - Communication Networks:
 - * Load balancing
 - * Queueing, Loss Networks → Ana
 - Network economics:
 - * Diffusions in social networks
 - * Control of epidemics
 - * Mean field games for security