# Ising models on power-law random graphs

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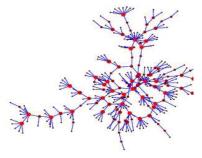
Where innovation starts

TU

## Introduction

There are many complex real-world networks, e.g.,

- Social networks (friendships, business relationships, sexual contacts, ...);
- Information networks (World Wide Web, citations, ...);
- Technological networks (Internet, airline routes, ...);
- Biological networks (protein interactions, neural networks,...).



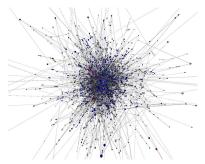
Sexual network Colorado Springs, USA (Potterat, et al., '02)



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Small part of the Internet (http://www.fractalus.com/ steve/stuff/ipmap/)



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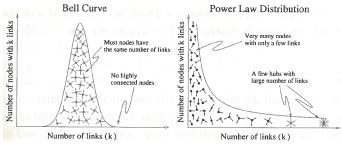


Yeast protein interaction network (Jeong, et al., '01)



#### Power-law behavior

Number of vertices with degree k is proportional to  $k^{-\tau}$ . Often,  $2 < \tau < 3$  (finite mean, infinite variance).



Barabási, Linked, '02

#### Small worlds Distances in the network are small



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# Ising model

Ising model: paradigm model in statistical physics for *cooperative behavior*.

When studied on complex networks it can model for example *opinion spreading* in society.

What are effects of *structure* of complex networks on *behavior* of Ising model?



## Power-law random graphs

In the *configuration model* a graph  $G_n = (V_n = [n], E_n)$  is constructed as follows.

- Let D have a certain distribution (the degree distribution);
- ► Assign D<sub>i</sub> half-edges to each vertex i ∈ [n], where D<sub>i</sub> are i.i.d. like D (Add one half-edge to last vertex when the total number of half-edges is odd);
- Attach first half-edge to another half-edge uniformly at random;
- Continue until all half-edges are connected.

Special attention to power-law degree sequences, i.e.,

$$\mathbb{P}[D \geq k] \leq ck^{-(\tau-1)}, \qquad \tau > 1.$$

## Local structure configuration model for $\tau > 2$

Start from random vertex i which has degree  $D_i$ .

Look at neighbors of vertex i, probability such a neighbor has degree k + 1 is approximately,

$$\frac{(k+1)\sum_{j\in[n]}\mathbb{1}_{\{D_j=k+1\}}}{\sum_{j\in[n]}D_j}$$



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Let *K* have distribution (the *forward degree* distribution),

$$\mathbb{P}[K = k] = \frac{(k+1)\mathbb{P}[D = k+1]}{\mathbb{E}[D]}$$

*Locally tree-like* structure: a branching process with offspring *D* in first generation and *K* in further generations. Also, *uniformly sparse*.

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## Definition of the Ising model

On a graph  $G_n$ , the *ferromagnetic Ising model* is given by the following Boltzmann distributions over  $\sigma \in \{-1, +1\}^n$ ,

$$\mu(\sigma) = \frac{1}{Z_n(\beta, B)} \exp\left\{\beta \sum_{(i,j)\in E_n} \sigma_i \sigma_j + \sum_{i\in[n]} B\sigma_i\right\},\,$$

#### where

- $\beta \ge 0$  is the inverse temperature;
- B is the external magnetic field;
- $Z_n(\beta, B)$  is a normalization factor (the *partition function*).



## **Critical temperature**

Define the *magnetization* on G<sub>n</sub> as

$$m_n(\beta, B) = \langle \frac{1}{n} \sum_{i=1}^n \sigma_i \rangle_\mu.$$

Then, the *spontaneous magnetization*,

$$M = \lim_{B \downarrow 0} \lim_{n \to \infty} m_n \begin{cases} = 0, & \beta < \beta_c; \\ > 0, & \beta > \beta_c. \end{cases}$$

The *critical inverse temperature*  $\beta_c$  is given by

 $\mathbb{E}[K](\tanh \beta_c) = 1.$ 

Note that, for  $\tau \in (2, 3)$ , we have  $\mathbb{E}[K] = \infty$ , so that  $\beta_c = 0$ .



Theorem (Dembo, Montanari, '08) For a locally tree-like and uniformly sparse graph sequence  $\{G_n\}_{n\geq 1}$  with  $\mathbb{E}[K] < \infty$ , the pressure per particle,

$$\psi_n(\beta, B) = \frac{1}{n} \log Z_n(\beta, B),$$

converges, for  $n \to \infty$ , to  $\varphi_h(\beta, B) \equiv \frac{\mathbb{E}[D]}{2} \log \cosh(\beta) - \frac{\mathbb{E}[D]}{2} \mathbb{E}[\log(1 + \tanh(\beta) \tanh(h_1) \tanh(h_2))]$ +  $\mathbb{E}\left[\log\left(e^{B}\prod_{i=1}^{D}\left\{1+\tanh(\beta)\tanh(h_{i})\right\}\right]\right]$  $+e^{-B}\prod_{i=1}^{D}\left\{1-\tanh(\beta)\tanh(h_i)\right\}\right)\right].$ TU/e Technische Universiteit / department of mathematics and computer science

### Theorem (DGvdH, '09)

Let  $\tau > 2$ . Then, in the configuration model, the pressure per particle,

$$\psi_n(\beta, B) = \frac{1}{n} \log Z_n(\beta, B),$$

converges almost surely, for  $n \to \infty$ , to

$$\varphi_{h}(\beta, B) \equiv \frac{\mathbb{E}[D]}{2} \log \cosh(\beta) - \frac{\mathbb{E}[D]}{2} \mathbb{E}[\log(1 + \tanh(\beta) \tanh(h_{1}) \tanh(h_{2}))] + \mathbb{E}\left[\log\left(e^{\beta} \prod_{i=1}^{D} \{1 + \tanh(\beta) \tanh(h_{i})\} + e^{-\beta} \prod_{i=1}^{D} \{1 - \tanh(\beta) \tanh(h_{i})\}\right)\right].$$



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## **Tree recursion**

## Proposition

Let  $K_t$  be i.i.d. like K and B > 0. Then, the recursion

$$h^{(t+1)} \stackrel{d}{=} B + \sum_{i=1}^{K_t} \operatorname{atanh}(\operatorname{tanh}(\beta) \operatorname{tanh}(h_i^{(t)})),$$

has a unique fixed point  $h_{\beta}^*$ .

Interpretation: the *effective field* of a vertex in a *tree* expressed in that of its neighbors.

Uniqueness shown by showing that effect of *boundary conditions* on generation *t* vanishes for  $t \to \infty$ .

This is done using *monotonicity* in  $\beta$  and *B* and *concavity* in *B* of the magnetization in the ferromagnetic Ising model.



$$= \lim_{\varepsilon \downarrow 0} \lim_{n \to \infty} \left[ \psi_n(\mathbf{0}, \mathbf{B}) + \int_0^\varepsilon \frac{\partial}{\partial \beta'} \psi_n(\beta', \mathbf{B}) \mathrm{d}\beta' + \int_\varepsilon^\beta \frac{\partial}{\partial \beta'} \psi_n(\beta', \mathbf{B}) \mathrm{d}\beta' \right]$$

$$= \varphi_h(\mathbf{0}, \mathbf{B}) + \mathbf{0} + \lim_{\varepsilon \downarrow \mathbf{0}} \int_{\varepsilon}^{\beta} \frac{\partial}{\partial \beta'} \varphi(\beta', \mathbf{B}) \mathrm{d}\beta'$$

 $= \varphi_h(\beta, \mathbf{B}).$ 



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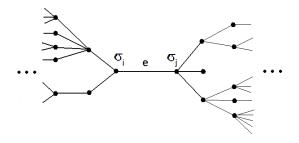
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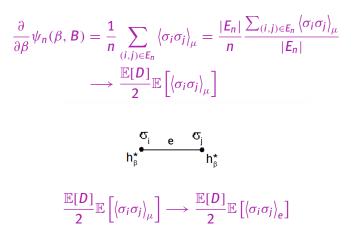
## **Derivative of pressure**

$$\frac{\partial}{\partial\beta}\psi_n(\beta,B) = \frac{1}{n}\sum_{(i,j)\in E_n} \langle\sigma_i\sigma_j\rangle_\mu = \frac{|E_n|}{n}\frac{\sum_{(i,j)\in E_n} \langle\sigma_i\sigma_j\rangle_\mu}{|E_n|}$$
$$\longrightarrow \frac{\mathbb{E}[D]}{2}\mathbb{E}\left[\langle\sigma_i\sigma_j\rangle_\mu\right]$$





## **Derivative of pressure**





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## **Derivative of** $\varphi$

$$\frac{\partial}{\partial\beta}\varphi_{h_{\beta}^{*}}(\beta,B) = \frac{\mathbb{E}[D]}{2}\mathbb{E}\left[\left\langle\sigma_{i}\sigma_{j}\right\rangle_{e}\right].$$

 $\varphi_h(\beta, B) = \frac{\mathbb{E}[D]}{2} \log \cosh(\beta) - \frac{\mathbb{E}[D]}{2} \mathbb{E}[\log(1 + \tanh(\beta) \tanh(h_1) \tanh(h_2))] \\ + \mathbb{E}\left[\log\left(e^B \prod_{i=1}^{D} \left\{1 + \tanh(\beta) \tanh(h_i)\right\} + e^{-B} \prod_{i=1}^{D} \left\{1 - \tanh(\beta) \tanh(h_i)\right\}\right)\right]$ 

- Show that we can ignore dependence of h<sup>\*</sup><sub>β</sub> on β;
   (*Interpolation* techniques. Split analysis into two parts, one for *small degrees* and one for *large degrees*)
- Compute the derivative with assuming β fixed in h<sup>\*</sup><sub>β</sub>.



## Distances in power-law random graphs

Let  $H_n$  be the graph distance between two *uniformly chosen connected* vertices in the configuration model. Then:

For  $\tau > 3$  and  $\mathbb{E}[K] > 1$  (vdH, Hooghiemstra, Van Mieghem, '05),

 $H_n \sim \log n$ ,

▶ For  $\tau \in (2, 3)$  (vdH, Hooghiemstra, Znamenski, '07),

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For  $\tau > 3$  and  $\tau \in (2, 3)$  similar results hold for the diameter of linear preferential attachment models (D, vdH, Hooghiemstra, '09).



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## **Critical exponents**

Predictions by physicists (Dorogovtsev, Goltsev, Mendes, '02 and Leone, Vázquez, Vespignani, Zecchina, '02).

Critical behavior for  $\beta \downarrow \beta_c$  of magnetization m, specific heat  $\delta C$  and susceptibility  $\chi$ .

	т	δC	χ
$\tau > 5$	$\sim (eta - eta_c)^{1/2}$	jump at $\beta_c$	$\sim (\beta - \beta_c)^{-1}$
$\tau \in (3, 5)$	$\sim (eta - eta_{c})^{1/( au - 3)}$	$\sim (eta - eta_{ m c})^{(5- au)/( au-3)}$	
$\tau \in (2,3)$	$\sim (eta - eta_c)^{1/(3- au)}$	$\sim (eta - eta_{c})^{( au-1)/(3- au)}$	$\sim (eta - eta_c)^1$

