Ising model on two connected Barabasi-Albert networks

Krzysztof Suchecki Janusz A. Hołyst

cond-mat/0603693, submitted to Phys. Rev. E

• Social interactions – often majority-driven



• Social interaction networks are often modular



• We represent simplest social interactions with Ising model on network



• And then study coupling of two networks (one larger modular network)



R

Regular networks

Hamiltonian for the Ising model, assuming the interaction constant is same for all pairs of spins.

$$H = -J\sum_{i,j} s_i s_j$$

Equation for mean spin (magnetization) in the mean-field approximation:

$$\langle s_i \rangle = \tanh \left| \beta \sum J_{ij} s_j \right| = \tanh \left(\beta J k \langle s_i \rangle \right)$$

k - coordination number (number of neighbors)

Scale-free networks $\langle s_i \rangle = \tanh\left(\beta \sum_{i} J_{ij} s_i\right) = \tanh\left(\beta J \sum_{i} \varepsilon_{ij} \langle s_i \rangle\right)$

Instead of coincidence matrix ε_{ij} we take the statistical probability of the spins being neighbors:

$$\left\langle \varepsilon_{ij} \right\rangle = \frac{k_i k_j}{E} = \frac{k_i k_j}{\langle k \rangle N}$$

We obtain following

$$\langle s_i \rangle = \tanh\left(\frac{\beta J k_i}{\langle k \rangle N} \sum_j k_j \langle s_j \rangle\right)$$

Scale-free networks

$$\langle s_i \rangle = \tanh\left(\frac{\beta J k_i}{\langle k \rangle N} \sum_j k_j \langle s_j \rangle\right)$$

Now we define average weighted spin S as:

$$S = \frac{1}{\langle k \rangle N} \sum_{i} k_i \langle S_i \rangle$$

so we can write above as

$$S = \frac{1}{\langle k \rangle N} \sum_{i} k_{i} \tanh(\beta J k_{i} S)$$

Scale-free networks

What does the weighted spin mean, and why weighted ?



k₁=6

 \mathbf{S}_1

k₂=2

 S_2

 $s_1 + s_2 = 0$ no order ?

 $s_1k_1 + s_2k_2 < 0$ order !

Scale-free networks

 $S = \frac{1}{\langle k \rangle N} \sum k_i \tanh(\beta J k_i S)$

Linear approximation

 $S = \frac{1}{\langle k \rangle N} \sum_{i} k_{i}^{2} \beta JS = \beta J \frac{\sum_{i} k_{i}^{2}}{\langle k \rangle N} S = \beta J \frac{\langle k^{2} \rangle}{\langle k \rangle} S$

B-A network

 $\beta J \frac{\langle k^2 \rangle}{\langle k \rangle} S = \beta J \frac{m}{2} \ln NS \longrightarrow T_c = J \frac{m}{2} \ln N$



Effective Tc versus m+ N for m = 5 and various N, averaged over up to 1000 samples.



Physica A 310 (2002) 260-266



www.elsevier.com/locate/physa

Ferromagnetic phase transition in Barabási–Albert networks

Agata Aleksiejuk^{a,*}, Janusz A. Hołyst^a, Dietrich Stauffer^b ^aFaculty of Physics, Warsaw University of Technology, Koszykowa 75, PL-00-662 Warsaw, Poland



Using linear approximation we investigate existence of nonzero solutions, that correspond to an ordered phase.

$$\langle s_{Ai} \rangle = \beta J_{AA} k_{AAi} \sum_{j} \frac{k_{Aj} \langle s_{Aj} \rangle}{E_{AA}} + \beta J_{BA} k_{ABi} \sum_{j} \frac{k_{BAj} \langle s_{Bj} \rangle}{E_{BA}}$$
$$\langle s_{Bi} \rangle = \beta J_{BB} k_{BBi} \sum_{j} \frac{k_{Bj} \langle s_{Bj} \rangle}{E_{BB}} + \beta J_{AB} k_{BAi} \sum_{j} \frac{k_{ABj} \langle s_{Aj} \rangle}{E_{AB}}$$

Similar to single scale-free network we introduce weighted spins S_{A} , S_B . Unfortunately we also need S_{AB} and S_{BA} .

 S_A - spin of network A weighted by k_{AA} (•) S_{BA} - spin of network B weighted by k_{BA} (•)

B



The state of the system can be written as vector:

 $\left(\begin{array}{c} S_{A} \\ S_{AB} \\ S_{BA} \\ S_{B} \end{array}\right)$

If we assume that the number of inter-network connections is proportional to the intra-network degree:

$$k_{AB} = p_A k_{AA}, \qquad k_{BA} = p_B k_{BB}$$

then we don not have to consider S_{AB} and S_{BA} anymore since they are proportional to S_A and S_B .

$$\begin{bmatrix} \Lambda_{AA} & \Lambda_{BA} \\ \Lambda_{AB} & \Lambda_{BB} \end{bmatrix} \begin{pmatrix} S_A \\ S_B \end{pmatrix} = \lambda \begin{pmatrix} S_A \\ S_B \end{pmatrix}$$

We have following eigenvalues λ of the matrix Λ :

$$A_{\pm} = \frac{\Lambda_{AA} + \Lambda_{BB} \pm \sqrt{(\Lambda_{AA} - \Lambda_{BB})^2 + 4\Lambda_{BA}\Lambda_{AB}}}{2}$$

 λ_{-c} : eigenvector $\begin{pmatrix} 1 \\ -c \end{pmatrix}$

The netwoks are ordered antiparalelly. T_C is lower than for separate networks.

Increasing inter-network connection strengths causes T_C to decrease, down to 0, when the inter-network connections are as dense as intra-network. λ_{+} : eigenvector $\begin{pmatrix} 1 \\ c_2 \end{pmatrix}$

The networks are ordered paralelly. T_C is higher than for separate networks.

Networks stabilize each other, similar to way the network stabilizes itself.

Numeric results

We have performed numeric simulations to find the critical temperatures T_{c+} and T_{c-} .



To find the critical temperature T_{c+} we calculate numerically the susceptibility $\chi = S_h - S_0$. Due to very highly fluctuating nature we make 30-point running average and fit parabolic curve.

Numeric results

We have performed numeric simulations to find the critical temperatures T_{c+} and T_{c-} .



Plot the total spin starting from antiparallel ordered state as a function of temperature.

Numerical results

Numerical simulations for two same B-A (N=5000, <k>=10, D=0) and various number of inter-network connections:



Inter-network link number

Questions

•What other dynamics can be investigated on connected networks and can they be described in the same way ?

•What other network topologies are worth investigating ?

Ref.: K. Suchecki, J.A. Holyst cond-mat/0603693