Emergence of communities in social networks

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### Emergence of communities in social networks?

Model of large social networks with focus on how communities emerge

Model should reproduce <u>characteristic properties</u> AND communities

Start from large-scale empirical social network

J.–P. Onnela, J. Saramäki, J. Hyvönen, G. Szabó, D. Lazer, K. Kaski, J. Kertész, and A.–L. Barabási, PNAS 104, 7332 (2007).

J. M. Kumpula, J.–P. Onnela, J. Saramäki, K. Kaski, and J. Kertész, Phys. Rev. Lett. 99, 228701 (2007).

### Overview

1. Social networks

2. Empirical social network

3. Modelling social networks

4. Conclusion

# Social networks

Social network paradigm in the social sciences: Social life consists of the flow and exchange of norms, values, ideas, and other social and cultural resources channelled through the social network

### Ø Perspective:

- Focus on very large networks
- Focus on statistical properties
- Complex networks & statistical mechanics



Photo from http://defiant.corban.edu/gtipton/net-fun/iceberg.html

# Social networks



- Ø Data from questionnaires; N ≈  $10^2$
- Scope of social interactions wide
- Strength based on recollection

New approach: 0

■ Electronic records of interactions; N ≈  $10^6$ 

- Scope of social interactions narrower
- Strength based on measurement





Constructed network is a proxy for the underlying social network 0

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# Constructing empirical network

#### 🔊 Data

One operator in a European country, 20% coverage
Aggregated from a period of 18 weeks
Over 7 million private mobile phone subscriptions
Voice calls within the operator
Require reciprocity of calls for a link
Quantify tie strength (link weight)



Aggregate call duration  $w_{ij}^D$ Total number of calls  $w_{ij}^N$ 



# About (social) network visualisation



Snowball sampling (distance!)Bulk nodes & surface nodes



Majority are surface nodesNeighbour visibility



## Local structure

Weak ties hypothesis\*: Relative overlap of two individual's friendship networks varies with the strength of their tie to one another

Define overlap O<sub>ij</sub> of edge (i,j) as the fraction of common neighbours

 Average overlap increases as a function of (cumulative) link weights

\* M. Granovetter, The strength of weak ties, AJS **78**, 1360 (1973)



- Probe the global role of links of different weight and local topology
- Approach of physicists (and children): Break to learn!
- Thresholding (percolation): Remove links based on their weight
- Control parameter f is the fraction of removed links
  - Initial network (f=0); isolated nodes (f=1)

Initial connected network (f=0), small sample ⇒ All links are intact, i.e. the network is in its initial stage



Decreasing weight thresholded network (f=0.8)  $\Rightarrow$  80% of the strongest links removed, weakest 20% remain



Initial connected network (f=0), small sample ⇒ All links are intact, i.e. the network is in its initial stage



Increasing weight thresholded network (f=0.8)  $\Rightarrow$  80% of the weakest links removed, strongest 20% remain



 Qualitative difference in the global role of weak and strong links
 Phase transition when weak ties are removed first f<sub>c</sub> (∞)≠1
 No phase transition when strong ties are removed first f<sub>c</sub> (∞)=1
 Suggests a point of division between weak and strong links (f<sub>c</sub>) W<sub>c</sub> = P<sup>-1</sup><sub>cum</sub> (0.80) ≈ 27 min



"globally connected" phase "disconnected islands" phase

Order parameter R<sub>LCC</sub> – Def: fraction of nodes in LCC Susceptibility S – Def: average cluster size (excl. LCC)

$$S = \sum_{s < s_{\text{max}}} n_s s^2 / \sum_{s < s_{\text{max}}} n_s s; \ \widetilde{S} = \sum_{s < s_{\text{max}}} n_s s^2 / N; \ C_i = t_i / 2k_i (k_i - 1)$$

# Summary of empirical study

Communities have mostly strong ties within (WTH)

Communities are interconnected mostly with weak ties (percolation)

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# Intro to modelling

- Social networks appear to have some "universal features"
- Can these features be reproduced with a simple microscopic model?
- Network sociology: How individual microscopic interactions translate into macroscopic social systems
- Statistical mechanics: How individual microscopic interactions translate into macroscopic (physical) systems

# Intro to modelling

### Internet & web => Simple rules work





# Intro to modelling

- A weighted model of social networks with focus on emergence of communities (mesoscopic structures) from microscopic rules
- Fixed number of nodes N
- Aim to reproduce characteristics features, no fitting to data
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- Regression models in sociology
- No claim for a grand unified theory of social networks

# Microscopic rules -> Mesoscopic structure







## Microscopic rules in the model

Local attachment (LA)





• Global (random) attachment (GA)  $k_i = 0 \implies P(i, j) = 1; w_{ij} = w_o = 1$  $k_i > 0 \implies P(i, j) = p_r; w_{ij} = w_o$ 

• Node deletion (ND)  $k_i > 0 \implies P(k_i = 0) = p_d$ 



### Microscopic rules in the model

### Local attachment (LA)

(1) Weighted local search / reinforcement  $P(i \rightarrow j) = w_{ij}/s_i$   $P(j \rightarrow k) = w_{jk}/(s_j - w_{ij})$   $w_{ij} \rightarrow w_{ij} + \delta$  $w_{jk} \rightarrow w_{jk} + \delta$ 



(2a) If (i,j,k) does not exist => Triangle formation  $P(i, j, k) = p_{\Delta}$   $w_{ik} = w_0 = 1$ (2b) If (i,j,k) exists => Triangle reinforcement  $w_{ik} \rightarrow w_{ik} + \delta$ 



### Microscopic rules in the model

Summary of the model

Weighted local search for new acquaintances
Reinforcement of popular links & Triangle formation
Unweighted global search for new acquaintances
Parameters

 $\begin{array}{ll} \delta & \quad \mbox{Free weight reinforcement parameter} \\ p_d = 10^{-3} & \quad \mbox{Sets the time scale of the model } \langle \tau_N \rangle = p_d^{-1} \\ p_r = 5 \times 10^{-4} & \quad \mbox{Global connections; Not sensitive} \\ \hline p_\Delta & \quad \mbox{Adjusted w.r.t. } \delta \mbox{ to keep } \langle k \rangle \mbox{ constant} \end{array}$ 

# Model mechanisms vs. sociology

Network sociology\*

Ocyclic closure

Second Exponential decay

Focal closure

Independent of distance

Sample window"

Model

Local attachment (LA)

Global attachment (GA)

Ø Node deletion (ND)

\* M. Kossinets et al., "Empirical Analysis of an Evolving Social Network", Science **311**, 88 (2006)

## **Basic characteristics**

(a) Fat-tailed degree distribution
(b) High clustering
(c) Assortative
(d) Small world

 $\begin{array}{l} \delta = 1 \\ \delta = 0.5 \\ \delta = 10^{-3} \\ \delta = 0 \end{array}$ 

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<u>▲</u> \*

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alizations of  $N = 5 \times 10^4$  networks. Values of  $\delta$  are 0 ( $\Box$ ),  $1 \times 10^{-3}$  (\*),  $1 \times 10^{-2}$  (>), 0.1 ( $\triangle$ ), 0.5 ( $\bigtriangledown$ ), and 1 ( $\circ$ ).



- Weak ties hypothesis (WTH)\*: implies weight-topology correlations: Ties within communities are strong, ties between communities are weak
- Sector States States
- $\circ$  Control parameter  $f \in [0,1]$
- ${oldsymbol{o}}$  Order parameter  $R_{
  m LCC}\in[0,1]$

<sup>\*</sup>M. Granovetter, "The Strength of Weak Ties", The American Journal of Sociology **78, 1360 (**1973)

Small 
$$\delta < 0.1$$
 
■  $\delta = 10^{-3}$ 
■  $\delta = 0$ 

 Network disintegrates at the same point for weak/strong link removal

Incompatible with WTH

■ Large δ > 0.1 ■ δ = 1 ■ δ = 0.5

- Network disintegrates at different points
- WTH compatible community structure



Weak go first Strong go first

### Communities by inspection

• Average number of links constant  $\langle L \rangle = N \langle k \rangle / 2$  => All changes in structure due to reorganisation of links

- Increasing  $\delta$  traps walks in communities, further enhancing trapping effect
  - => Clear communities
- Triangles accumulate weight and act as nuclei for communities



## Communities by k-clique method

- Ose k-clique algorithm / definition for communities<sup>\*</sup>
  Focus on 4-cliques (smallest non-trivial cliques)
  Relative largest community size R<sub>k=4</sub> ∈ [0, 1]
  Average community size (excl. largest) ⟨n⟩
- ${\it ilde o}$  Observe clique percolation through the system for small  $\delta$

 ${\it {\circ}}$  Increasing  $\delta$  leads to condensation of communities



### Is community size distribution stable?

 $\odot$  Consider community k with size N<sub>k</sub>

In the large  $\delta$  regime, most local random walks remain in the initial community, resulting in stable distribution

$$\frac{\mathrm{d}N_k}{\mathrm{d}t} = -p_d N_k + p_d N \frac{N_k}{N} = 0$$

Community formation happens in transient state

A triesgleeneringmulating weight acts as a nucleus for the enveriging ecommunity during subsequent LA steps

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## Conclusion

Local coupling between network topology and tie strengths (WTH)
Weak ties (PT) are qualitatively different from strong ties (no PT)
Model: essential characteristics & local & global properties

Need focal & cyclic closure & sufficient reinforcement of connections

 Communities result from initial structural fluctuations that become amplified by repeated application of the microscopic processes

## References

- J.-P. Onnela, J. Saramäki, J. Hyvönen, G. Szabó, D. Lazer, K. Kaski, J. Kertész, and A.-L. Barabási, "Structure and tie strengths in mobile communication networks", PNAS 104, 7332 (2007).
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- See also Science 314, 914 (2006).
- See http://www.physics.ox.ac.uk/users/Onnela/

### THANK YOU!