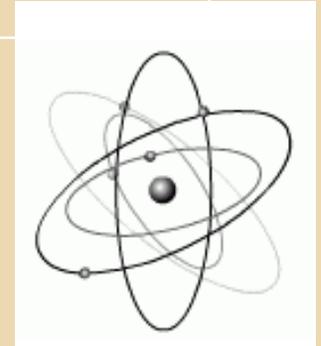


# Looking for communities in networks



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in cooperation with

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Summer School on Socio-Econo-Physics 2007 in Windberg

# Overview

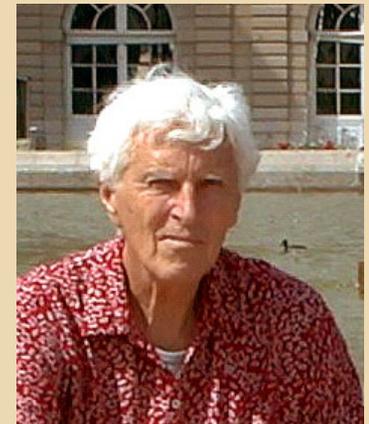
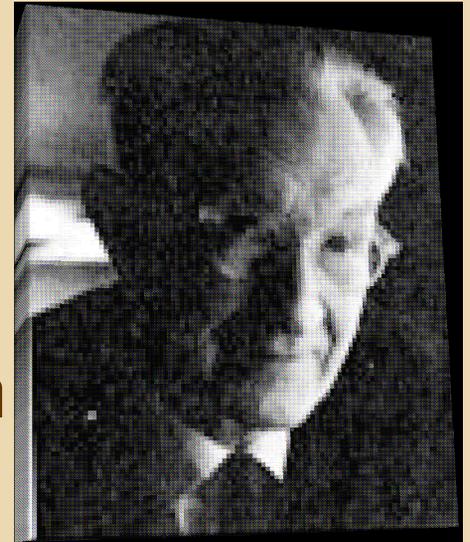
- The task
- A case study – the Southern Women
- The modularity after Newman
- A social problem – the Heider balance
- A continuous approach

Review:

S. Boccaletti et al.,

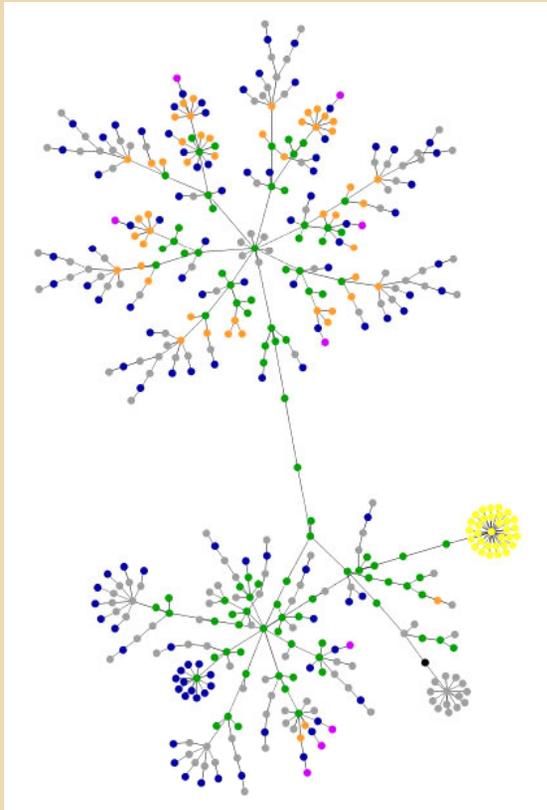
*Complex networks: Structure and dynamics*

Physics Reports 424 (2006) 175, section 7.1



# The task

- Graph representation of a network



Jasmine's blog

$N$  nodes

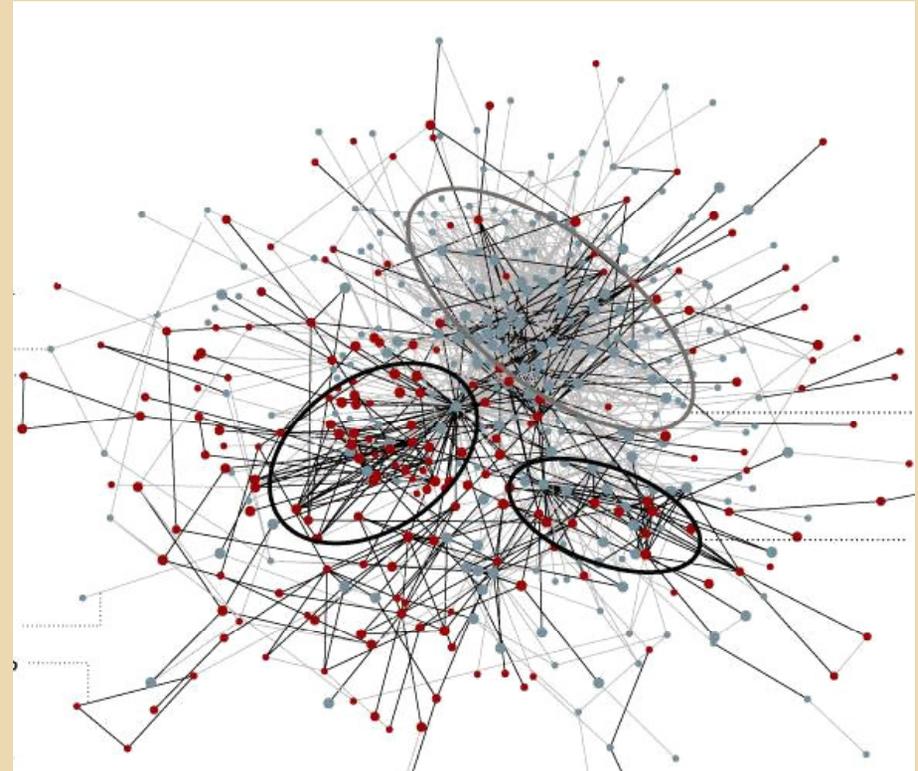
$A(i,k)$ – connectivity matrix

$$A(i,k) = 0, 1 \quad \text{or}$$

$$A(i,k) \in \mathcal{R} \quad (\text{weighted networks})$$

# The task

- The communities:
  - subsets of nodes, internally connected **more** than with other parts of the network



Cornell > D.Constantine, NYT

Our aim is to find **separated** communities.

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# The Southern Women, Natchez 1935



L. C. Freeman, *Finding social groups: a meta-analysis of the southern women data*, in Ronald Breiger, Kathleen Carley and Philippa Pattison (eds.) *Dynamic Social Network Modeling and Analysis*. Washington, D.C.:The National Academies Press, 2003.

# Participation of the Southern Women in Events

NAMES OF PARTICIPANTS OF GROUP I	CODE NUMBERS AND DATES OF SOCIAL EVENTS REPORTED IN <i>Old City Herald</i>													
	(1) 6/27	(2) 3/2	(3) 4/12	(4) 9/26	(5) 2/25	(6) 5/19	(7) 3/15	(8) 9/16	(9) 4/8	(10) 6/10	(11) 2/23	(12) 4/7	(13) 11/21	(14) 8/3
1. Mrs. Evelyn Jefferson.....	X	X	X	X	X	X	...	X	X	...	...	...	...	...
2. Miss Laura Mandeville.....	X	X	X	...	X	X	X	X	...	...	...	...	...	...
3. Miss Theresa Anderson.....	...	X	X	X	X	X	X	X	X	...	...	...	...	...
4. Miss Brenda Rogers.....	X	...	X	X	X	X	X	X	...	...	...	...	...	...
5. Miss Charlotte McDowd.....	...	...	X	X	X	...	X	...	...	...	...	...	...	...
6. Miss Frances Anderson.....	...	...	X	...	X	X	...	X	...	...	...	...	...	...
7. Miss Eleanor Nye.....	...	...	...	...	X	X	X	X	...	...	...	...	...	...
8. Miss Pearl Oglethorpe.....	...	...	...	...	...	X	...	X	X	...	...	...	...	...
9. Miss Ruth DeSand.....	...	...	...	...	X	...	X	X	X	...	...	...	...	...
10. Miss Verne Sanderson.....	...	...	...	...	...	...	X	X	X	...	...	X	...	...
11. Miss Myra Liddell.....	...	...	...	...	...	...	...	X	X	X	...	X	...	...
12. Miss Katherine Rogers.....	...	...	...	...	...	...	...	X	X	X	...	X	X	X
13. Mrs. Sylvia Avondale.....	...	...	...	...	...	...	X	X	X	X	...	X	X	X
14. Mrs. Nora Fayette.....	...	...	...	...	...	X	X	...	X	X	X	X	X	X
15. Mrs. Helen Lloyd.....	...	...	...	...	...	...	X	X	...	X	X	X	...	...
16. Mrs. Dorothy Murchison.....	...	...	...	...	...	...	...	X	X	...	...	...	...	...
17. Mrs. Olivia Carleton.....	...	...	...	...	...	...	...	...	X	...	X	...	...	...
18. Mrs. Flora Price.....	...	...	...	...	...	...	...	...	X	...	X	...	...	...



# Comparison of 21 methods

	<b>Code</b>	<b>Analysis</b>	<b>Closeness to the Matching Criterion</b>
1	DGG41	Davis, Gardner and Gardner, Ethnography	0.920
2	HOM50	Homans, Intuition	0.854
3	P&C72	Phillips and Conviser, Information Theory	0.968
4	BGR74	Breiger, Algebra	0.933
5	BBA75	Breiger, Boorman and Arable, CONCOR	0.927
6	BCH78	Bonacich, Boolean Algebra	0.841
7	DOR79	Doreian, Algebraic Topology	0.923
8	BCH91	Bonacich, Correspondence Analysis	0.968
9	FRE92	Freeman, G-Transitivity	0.926
10	E&B93	Everett and Borgatti, Regular Coloring	0.916
11	FR193	Freeman, Genetic Algorithm 1	0.968
12	FR293	Freeman, Genetic Algorithm 2	0.842
13	FW193	Freeman and White, Galois Lattice	0.917
14	FW293	Freeman and White, Galois Sub-Lattice	0.954
15	BE197	Borgatti and Everett, Bi-Clique	0.916
16	BE297	Borgatti and Everett, Taboo Search	0.968
17	BE397	Borgatti and Everett, Genetic Algorithm	0.968
18	S&F99	Skvoretz and Faust, p* Model	0.957
19	ROB00	Roberts, SVD with Normalization	0.968
20	OSB00	Osborn, VERI Algorithm	0.543
21	NEW01	Newman, Weighted Co-Attendance	0.932

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# Basic tool - the modularity

$$Q = \frac{1}{w} \sum_{ij} \left( A(i, j) - \frac{k_i k_j}{w} \right) \delta(c_i, c_j)$$

where

$$w = \sum_{ij} A(i, j)$$

$$k_i = \sum_j A(i, j)$$

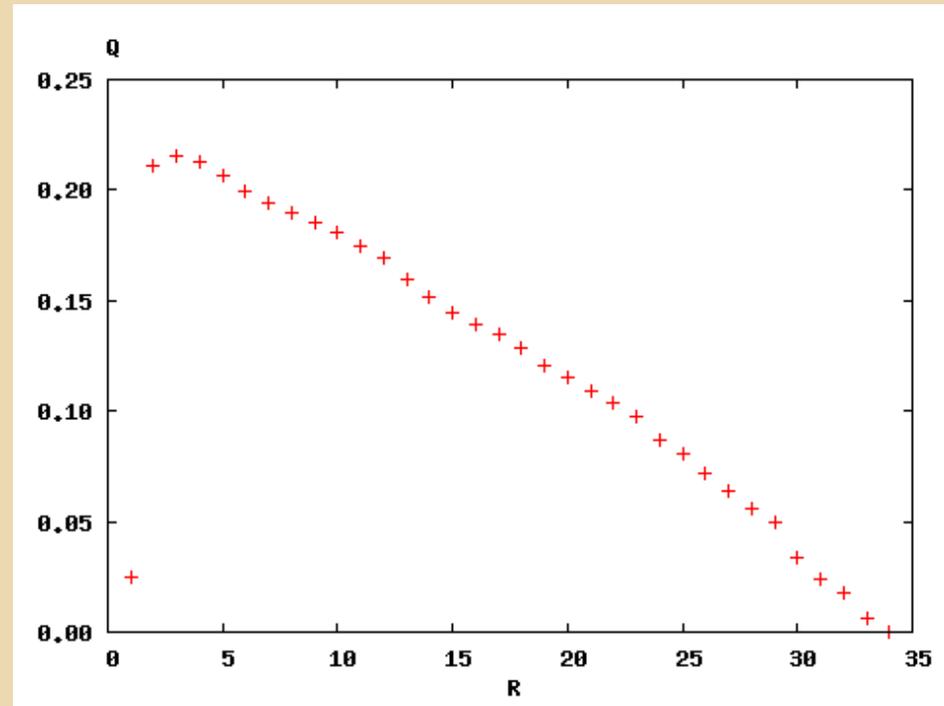
[M. E. J. Newman, *Analysis of weighted networks*, Phys. Rev. E 70, 056131 (2004)]

# Special cases

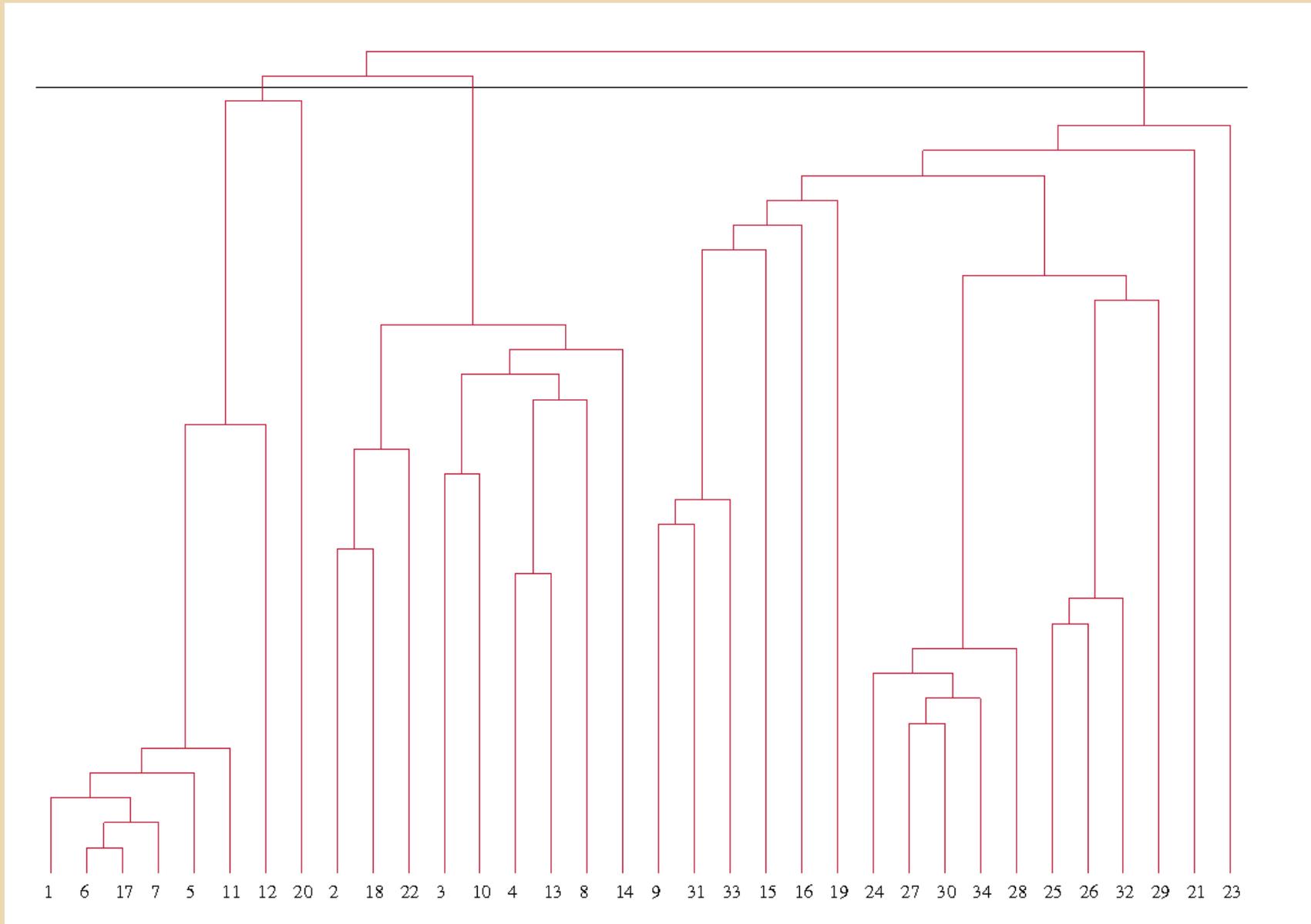
- One fully connected cluster of  $N$  nodes plus one separated node:  $Q = 1/N$ . When the node is added to the cluster,  $Q=0$ .
- $R$  separated fully connected clusters, each of  $N$  nodes:  $Q=1-(N-1)/(RN) \approx 1$  if  $R \gg 1$ . When one fully connected cluster is formed,  $Q=1/(RN) \approx 0$ .

# Algorithm

- For a given set of data  $(N, \{A(i,k)\})$  we have to construct a new connectivity matrix  $\{c_{ik}=0, 1\}$ , which reveals the communities.
- start from  $c_{ik}=0$  for all  $i,k$ .
- find the pair  $(n,m)$  such that when we put  $c_{nm}=1$ ,  $Q$  increases most.
- repeat the search of such pairs, until  $Q$  starts to decrease. We accept the division where  $Q$  has a maximum.



# Results – a dendrogram



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# The Heider balance

- Let us transform the connectivity matrix as:

1  $\rightarrow$  1            (friendly relation)

0  $\rightarrow$  -1            (hostile relation)

- Nodes = group members.
- In the balanced state, some rules are obeyed:

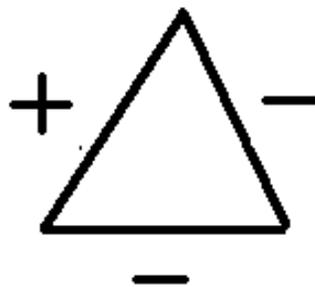
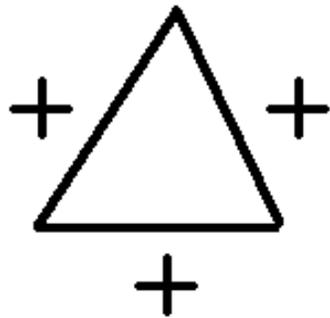
Friend of my friend is my friend     $A(i,k)>0$  ,  $A(k,n)>0 \Rightarrow A(i,n)>0$

Enemy of my friend is my enemy     $A(i,k)<0$  ,  $A(k,n)>0 \Rightarrow A(i,n)<0$

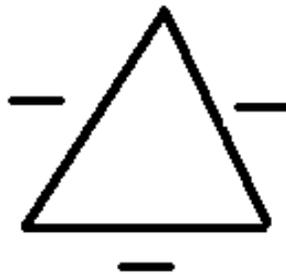
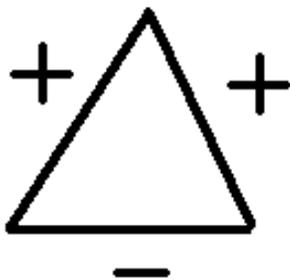
Friend of my enemy is my enemy     $A(i,k)>0$  ,  $A(k,n)<0 \Rightarrow A(i,n)<0$

Enemy of my enemy is my friend     $A(i,k)<0$  ,  $A(k,n)<0 \Rightarrow A(i,n)>0$

- If the state is not balanced, some members suffer the cognitive dissonance.



balanced  
triads

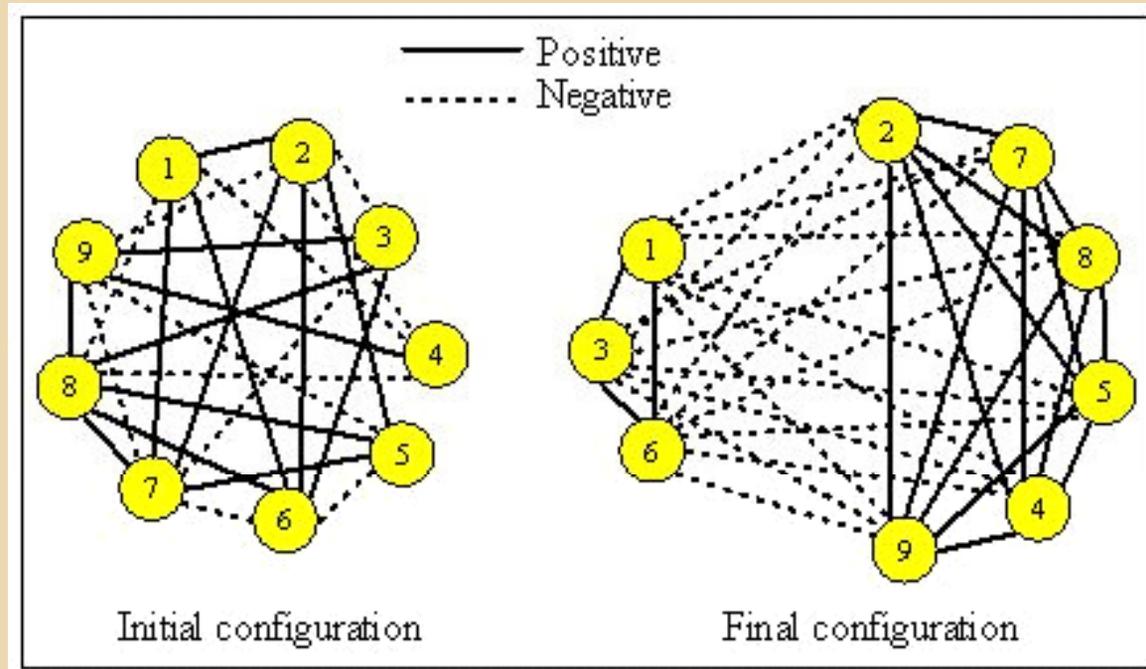


unbalanced  
triads

- They try to get the balanced state, changing some of their relations.

# The Harary theorem

- At the balanced state, the group is divided into two mutually hostile subgroups. The relations within both subgroups are friendly.



[Z. Wang, W. Thorngate, *Sentiment and social mitosis: Implications of Heider's Balance Theory*, J. of Artificial Society and Social Simulation vol. 6, no. 3 (2003)]

# Problems with the dynamics

- **Sociological:**

Relations are more subtle than just friendly or hostile; more possibilities should be included.

- **Computational:**

The final balanced state can depend on the order of modified relations.

# A continuous dynamics

- The relations are described by real numbers  $A(i,k) \in (-R,R)$
- Their time evolution is determined by differential equations

$$\frac{dA(i,k)}{dt} = \left[ 1 - \frac{A(i,k)^2}{R^2} \right] \sum_n A(i,n) A(n,k)$$

The results **agree** with the best six methods, reported by Freeman.

[More details in KK, *Some recent attempts to simulate the Heider balance problem*, Computing in Science and Engineering, July/August 2007, p.80.]

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# Looking for communities in general– - the continuous approach

- In the Heider balance problem, the differential equations were motivated by the specific nature of the cognitive dissonance.
- In a general approach, for each pair we ask: **do other connections confirm that this pair belongs to the same set?**

# The continuous approach

- The differential equations are

$$\frac{dA(i,k)}{dt} = \Theta(A(i,k))\Theta(1-A(i,k)) \sum_n [A(i,n)A(n,k) - b]$$

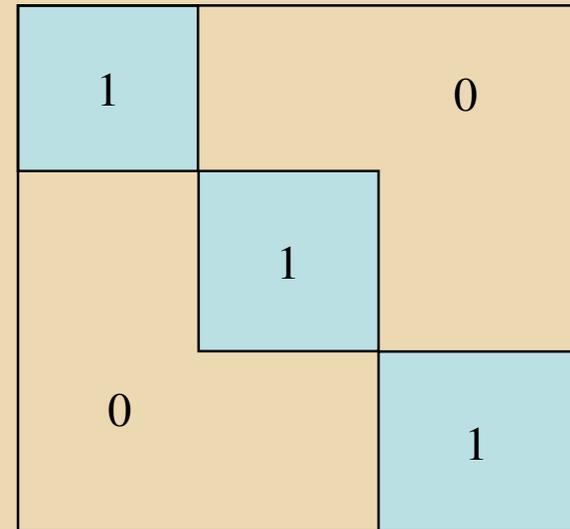
The product of the step functions ensures that once  $A(i,k)=0$  or  $1$ , it does not change any more.

Once the product  $A(i,n)A(n,k) > b$ , the connection of sites  $i,k$  via site  $n$  proves that  $i,n$  belong to the same cluster and so do  $n,k$ . Then  $A(i,k)$  should increase.

(more details in M. Krawczyk and K. K., [arXiv:0709.0923](https://arxiv.org/abs/0709.0923) )

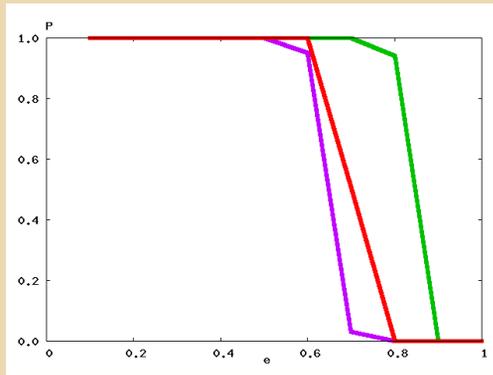
# The continuous approach – a test

- For separated and fully connected clusters, the connectivity matrix has a block form

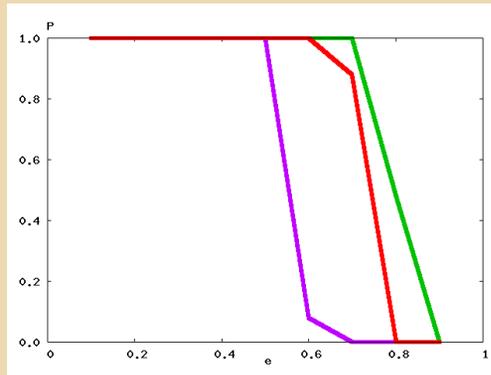


We add a noise: random numbers with the uniform distribution  $(0, e)$  are added to zeros and subtracted from units. The task is to reproduce the original structure.

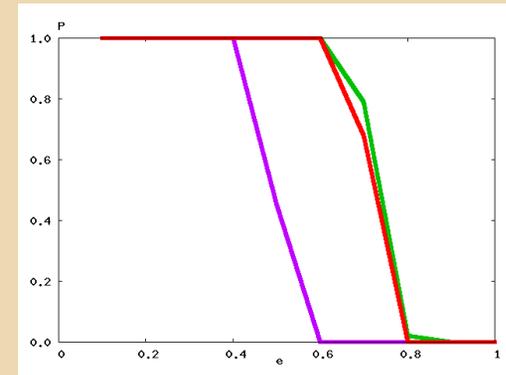
# The continuous approach – a test



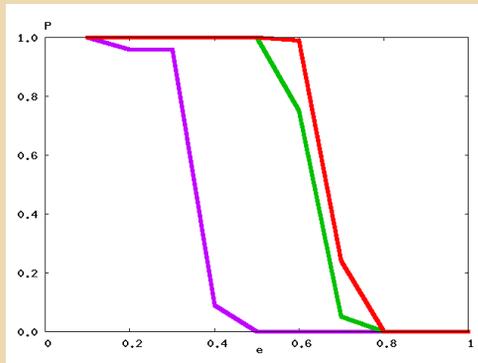
R=2



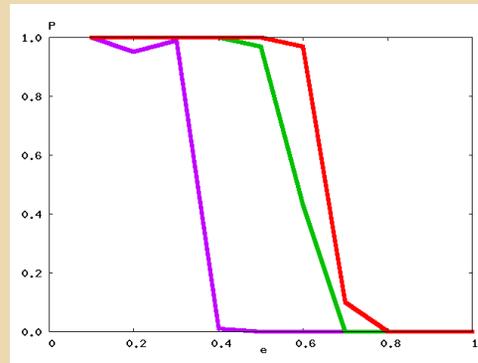
R=3



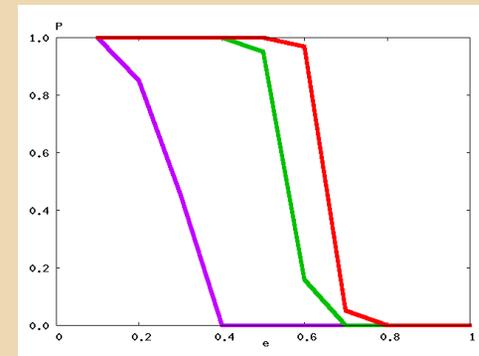
R=4



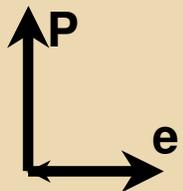
R=6



R=7



R=8



P – probability that the initial structure is reproduced  
 e – upper limit of noise

NR=60÷72, #=100, b=1/4,

— modularity

— random walk

— continuous

# Final remarks

- The method is deterministic. It produces some results even for a purely random graph.
- Direct comparisons with social experiments (Southern women, Zachary karate club) can be dubious.

Thank you

# Control questions

- To calculate the modularity, one takes into account:
  - A. Only existing connections
  - B. Only a proposed division into clusters
  - C. Both A and B
- Differential equations raise difficulties because
  - A. high numerical precision is needed
  - B. the order of modified relations does matter