

# Strength of Weak Ties and Community Structure in Networks

CS224W: Social and Information Network Analysis  
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<http://cs224w.stanford.edu>



# Networks: Flow of information

- How information flows through the network?
- How different nodes can play structurally distinct process in roles in this process?
- How different links (short range vs. long range) play different roles in diffusion?

# Strength of weak ties

[Granovetter '73]

- How people find out about new jobs?
  - Mark Granovetter, part of his PhD in 1960s
  - People find the information through personal contacts
- But: Contacts were often **acquaintances** rather than close friends
  - This is surprising:
    - One would expect your friends to help you out more than casual acquaintances when you are between the jobs
- Why is it that **distance acquaintances are most helpful?**

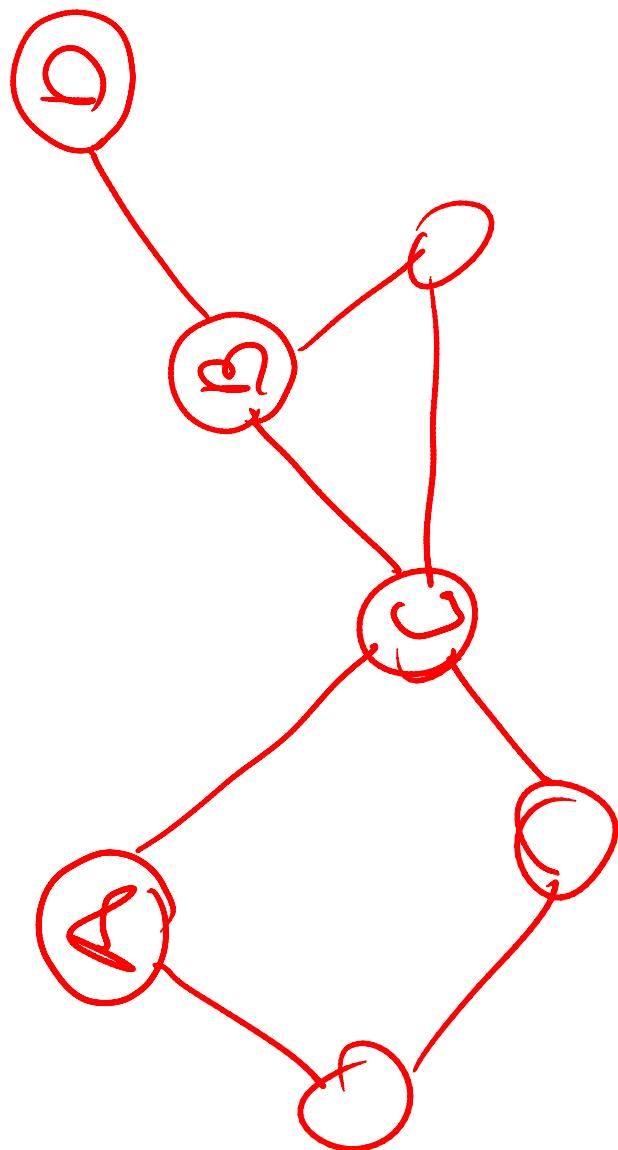
# Granovetter's answer

[Granovetter '73]

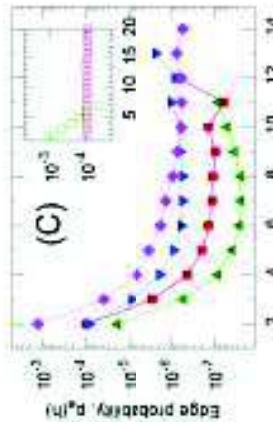
- Two perspectives on **friendships**:
  - **Structural**:
    - Friendships span different portions of the network
  - **Interpersonal**:
    - Friendship between two people is either strong or weak

# Triadic closure

- Which edge is more likely A-B or A-D?



- **Triadic closure:** If two people in a network have a friend in common there is an increased likelihood they will become friends themselves

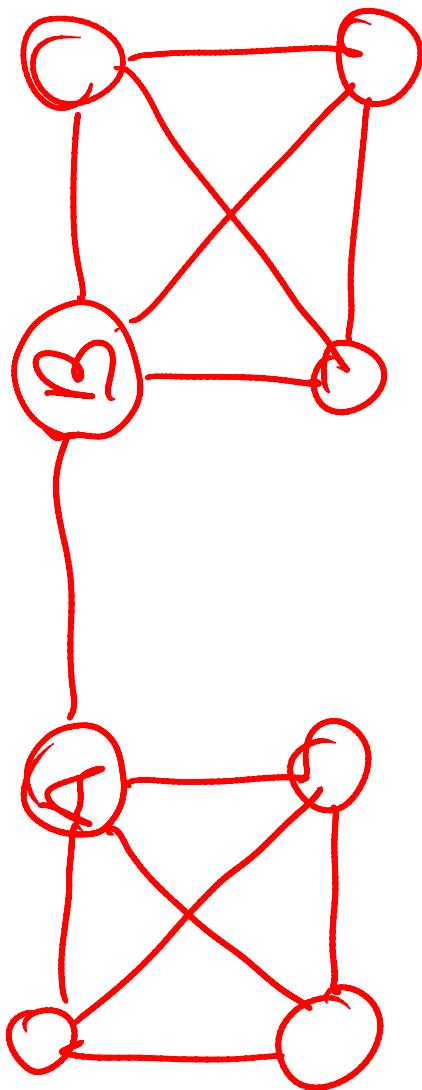


# Triadic closure

- Triadic closure == High clustering coefficient
- Reasons for triadic closure:
  - If B and C have a friend A in common, then:
    - B is more likely to meet C
      - (since they both spend time with A)
      - B and C trust each other
        - (since they have a friend in common)
    - A has incentive to bring B and C together
      - (as it is hard for A to maintain two disjoint relationships)
  - Empirical study by Bearman and Moody:
    - Teenage girls with low clustering coefficient are more likely to contemplate suicide

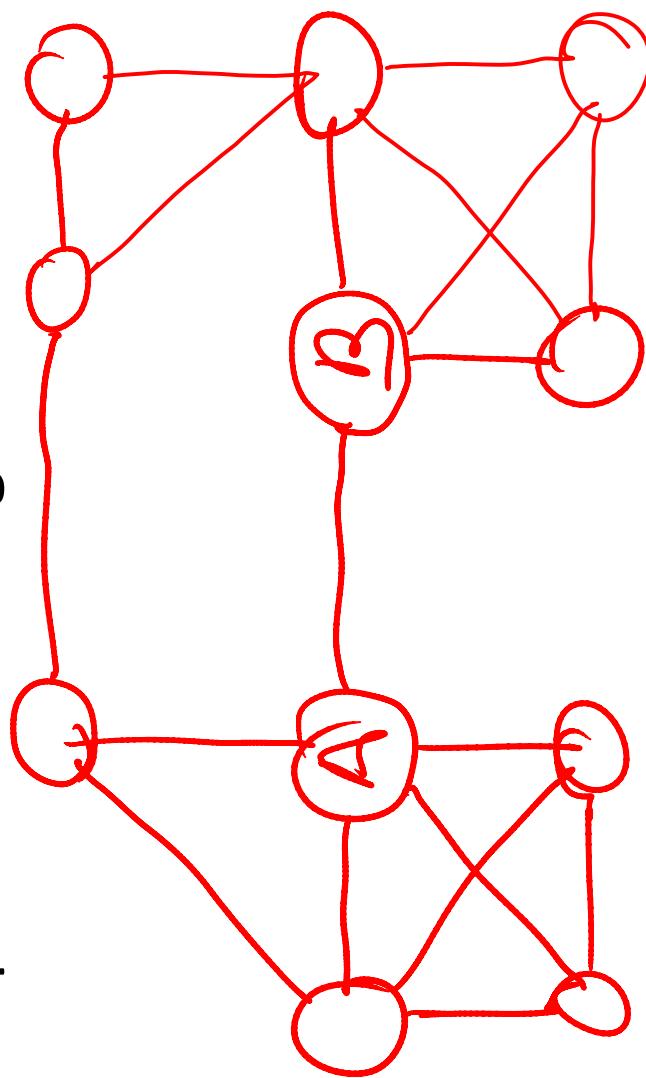
# Bridges and Local Bridges

- Edge  $(A, B)$  is a **bridge** if deleting it would make A and B be in two separate connected components.



# Bridges and Local Bridges

- Edge  $(A, B)$  is a **local bridge** if A and B have no friends in common
- **Span** of a local bridge is the distance of the edge endpoints if the edge is deleted



(local bridges with long span are like real bridges)

# Strong Triadic Closure

- **Links in networks have strength:**
  - Friendship
  - Communication
- We characterize links as either **Strong** (friends) or **Weak** (acquaintances)
- **Def: Strong Triadic Closure Property:**  
If A has **strong** links to B and C, then there must be a link (B,C) (that can be **strong** or **weak**)

# Local Bridges and Weak ties

- **Claim:** If node A satisfies Strong Triadic Closure and is involved in at least two **strong** ties, then any **local bridge** adjacent to A must be a **weak** tie.
- **Proof by contradiction:**
  - A satisfies Strong Triadic Closure
  - Let A-B be local bridge and a **strong** tie
  - Then B-C must exist because of Strong Triadic Closure
  - But then (A,B) is **not** a **bridge**

# Summary of what we just did

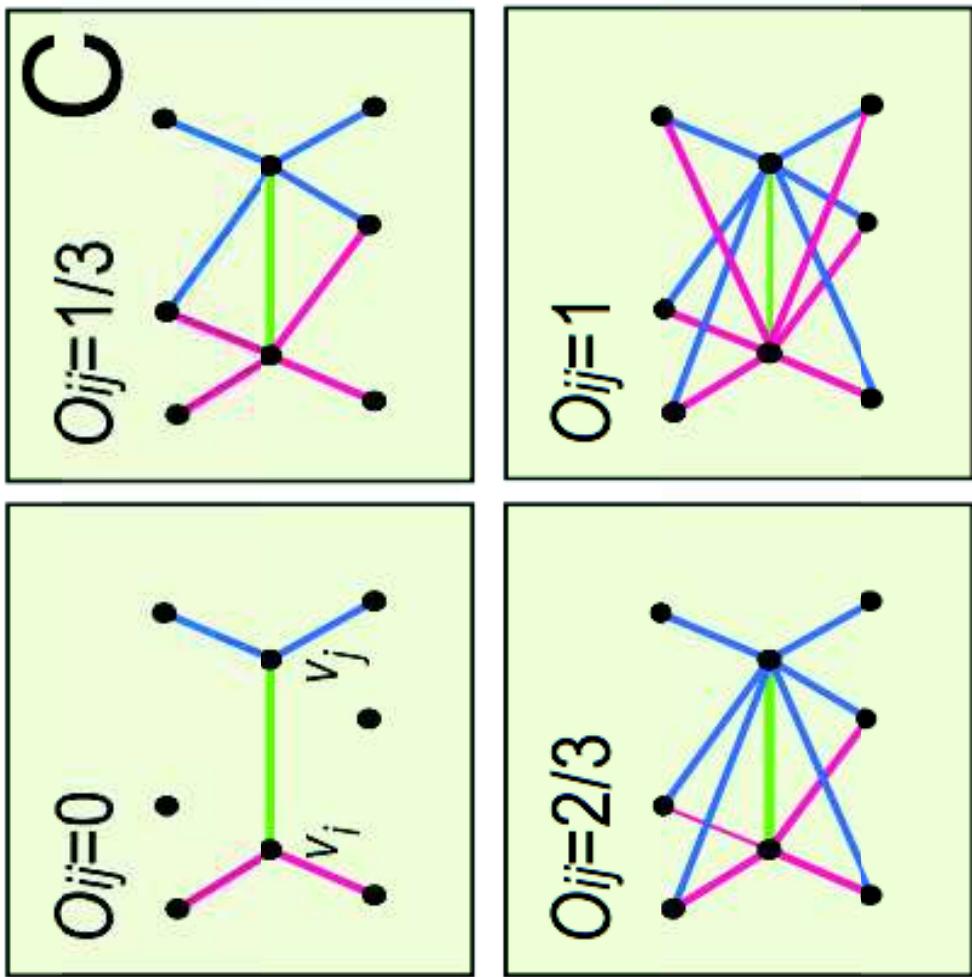
- **Defined Local Bridges:**
  - Edges not in triangles
- **Set two types of edges:**
  - **Strong and Weak Ties**
- **Defined Strong Triadic Closure:**
  - Two strong ties imply a third edge
- **→ Local bridges are weak ties**

# Tie strength in real data

- For many years the Granovetter's theory was not tested
- But, today we have large who-talks-to-whom graphs:
  - Email, Messenger, Cell phones, Facebook
- Onnela et al. 2007:
  - Cell-phone network of 20% of country's population

# Neighborhood Overlap

- Overlap:  
$$O_{ij} = \frac{n(i) \cap n(j)}{n(i) \cup n(j)}$$
  - $n(i)$  ... set of neighbors of A

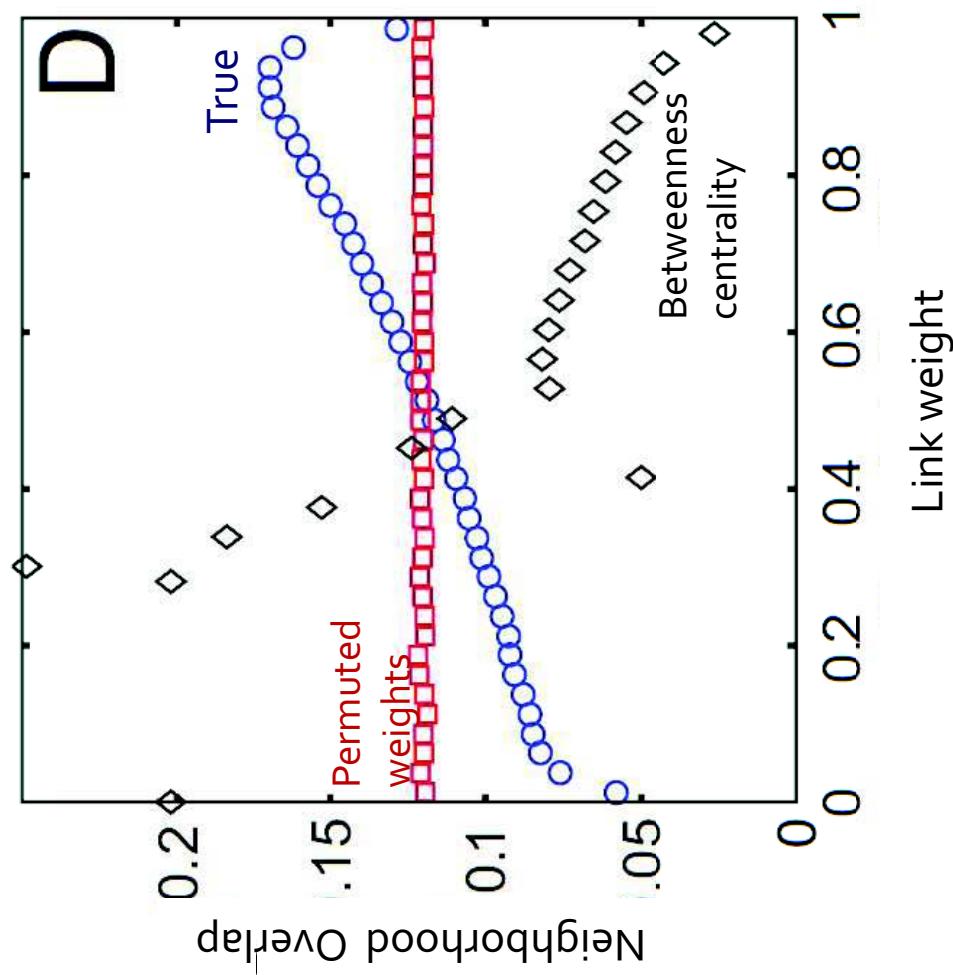


- Overlap = 0  
when an edge is a local bridge

# Mobile phones: Overlap vs. Weight

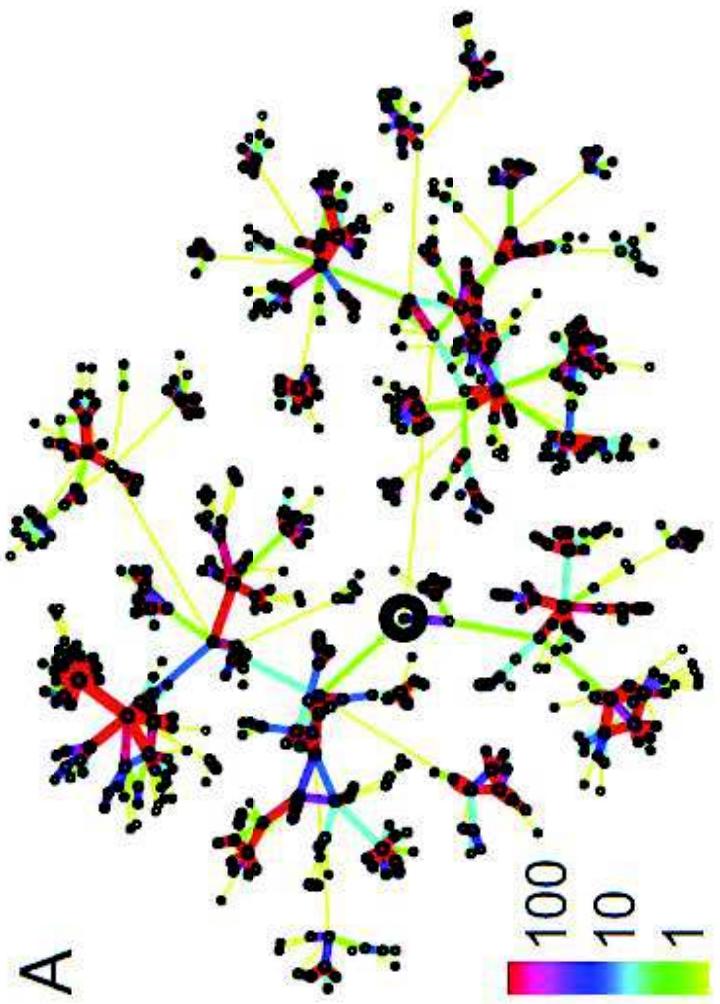
[Onnela et al. '07]

- **Permuted weights:** Keep the structure but randomly reassign edge weights
- **Betweenness centrality:** Number of shortest paths going through an edge



# Real network tie strengths

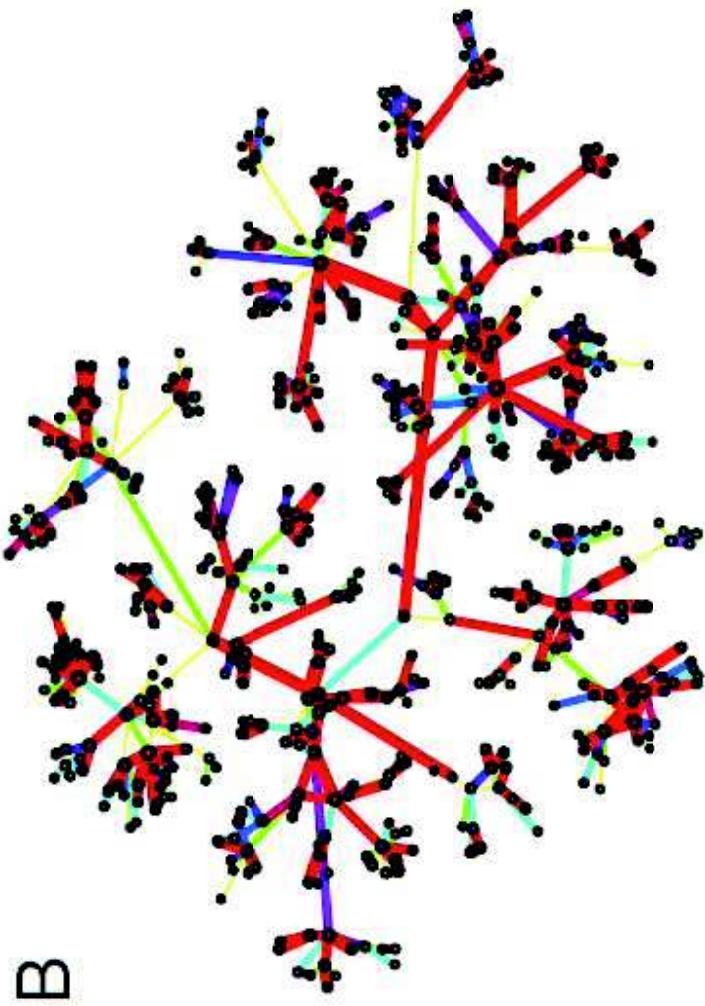
[Onnela et al. '07]



- Real edge strengths in mobile call graph

# Permuted tie strengths

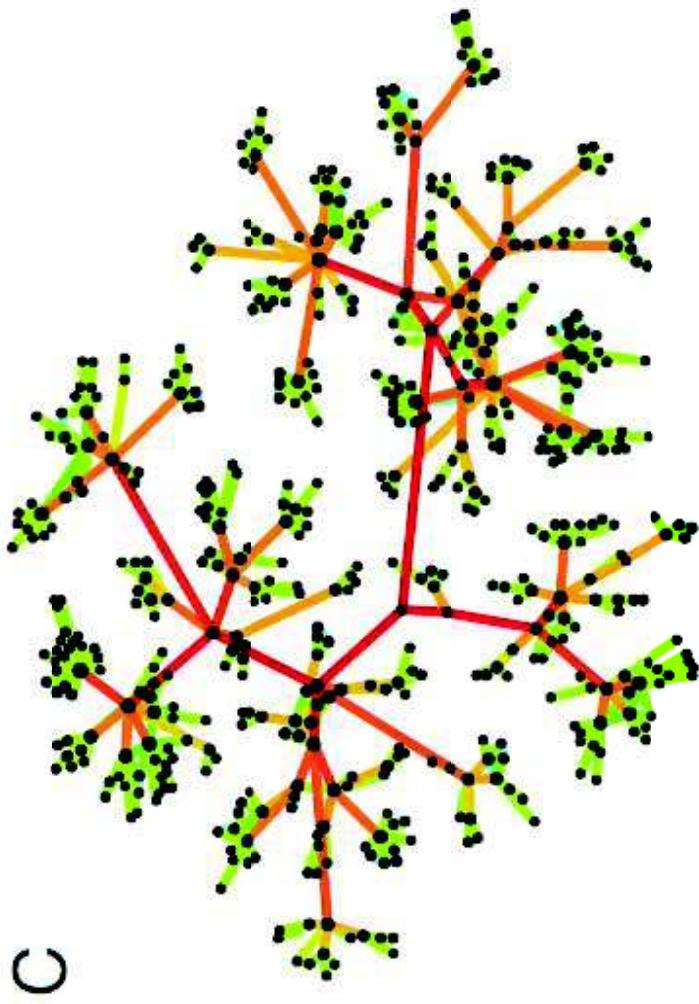
[Onnela et al. '07]



- Same network, same set of edge strengths
- But now **strengths are randomly shuffled** over the edges

# Edge betweenness centrality

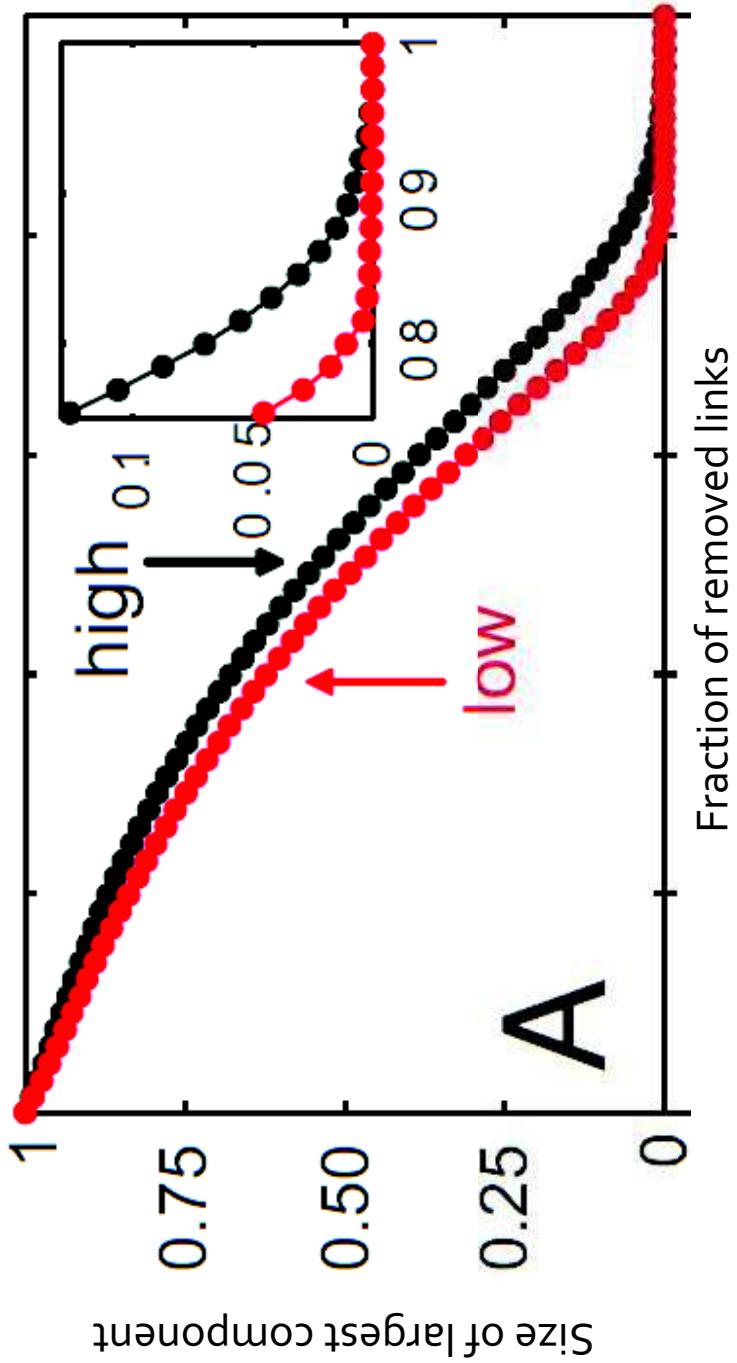
[Onnela et al. '07]



- Edges labeled based on **betweenness centrality** (number of shortest paths going through an edge)

# Link removal: Weight

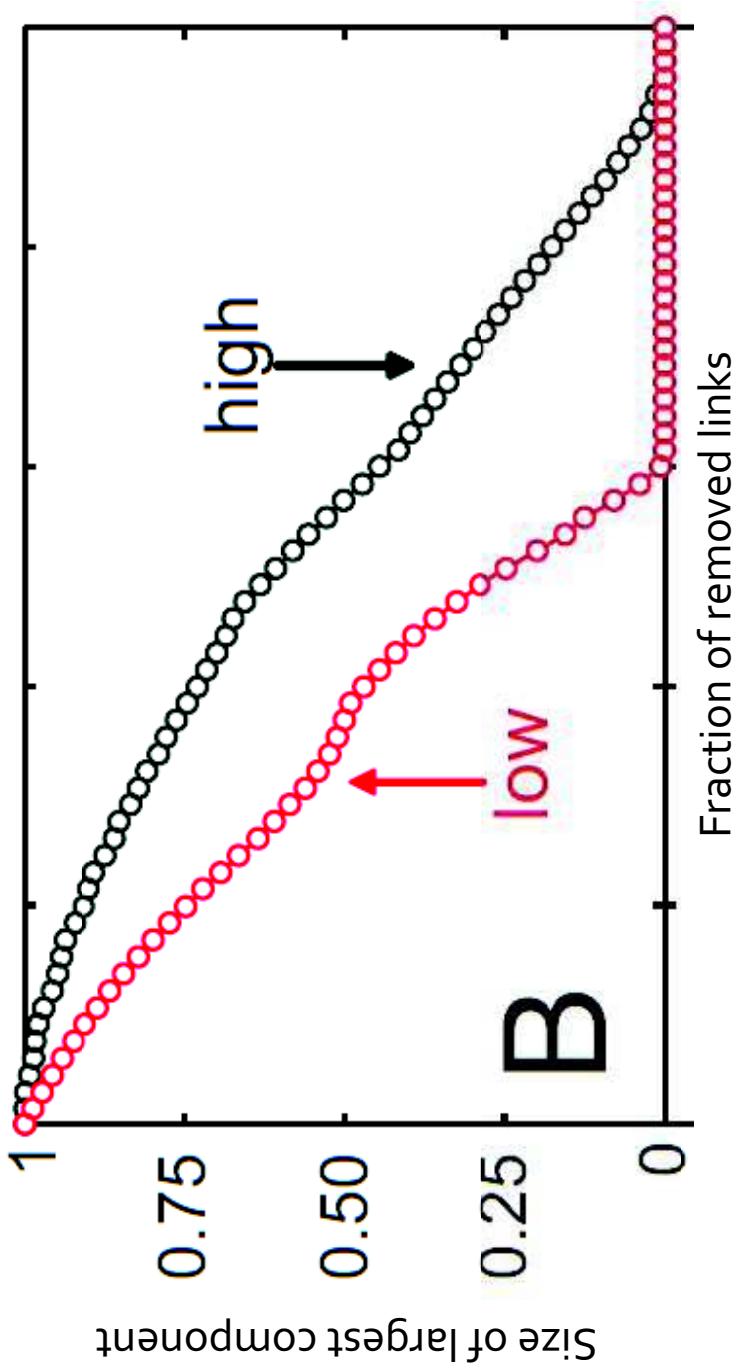
[Onnela et al. '07]



- Removing links based on **strength** (# conversations)
  - Low to high
  - High to low

# Link removal: Overlap

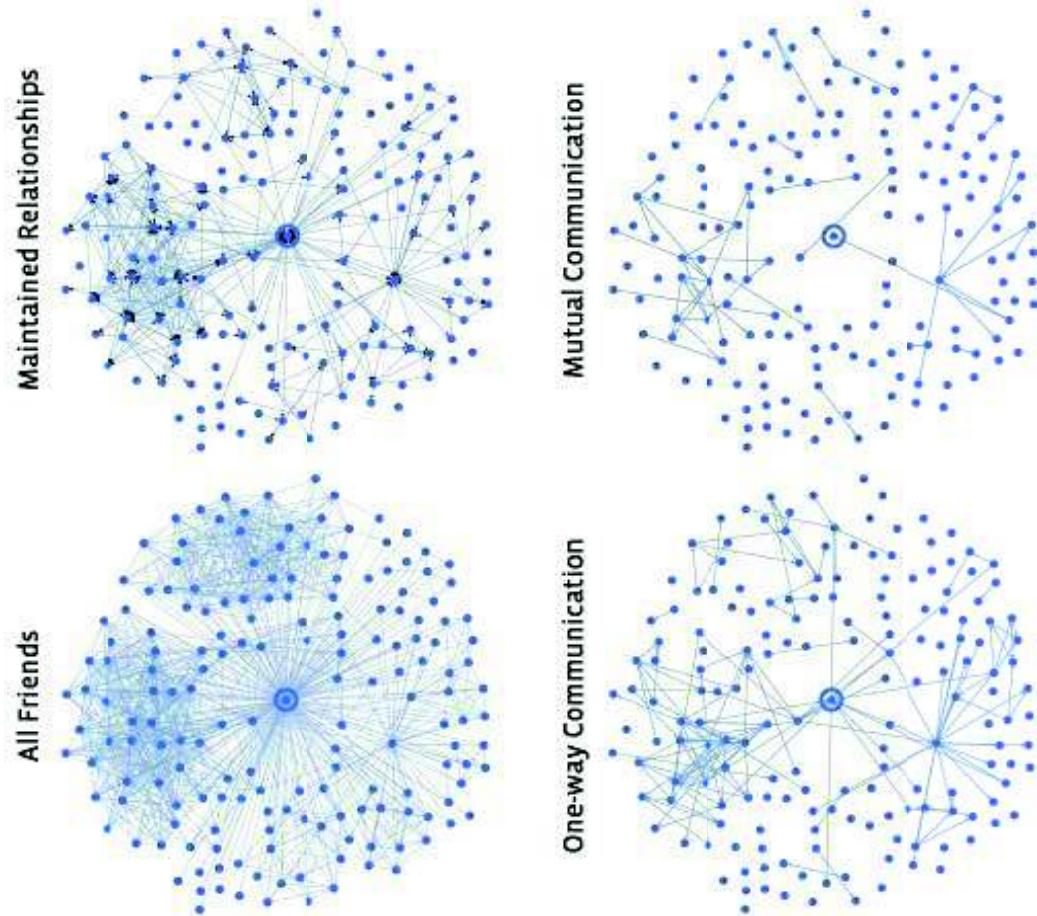
[Onnela et al. '07]



- Removing links based on **overlap**
  - Low to high
  - High to low

# Another example: Facebook

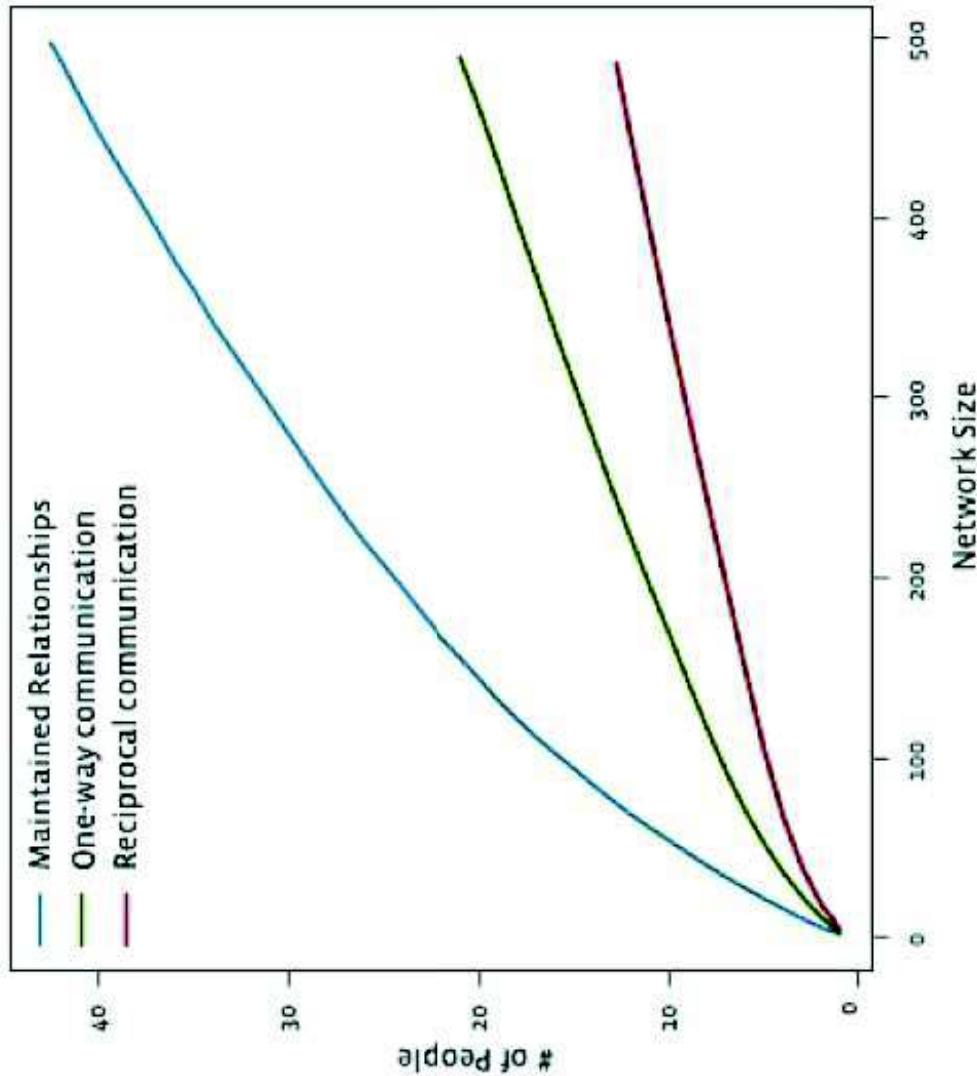
[Marlow et al. '09]



# Facebook: Number of ties

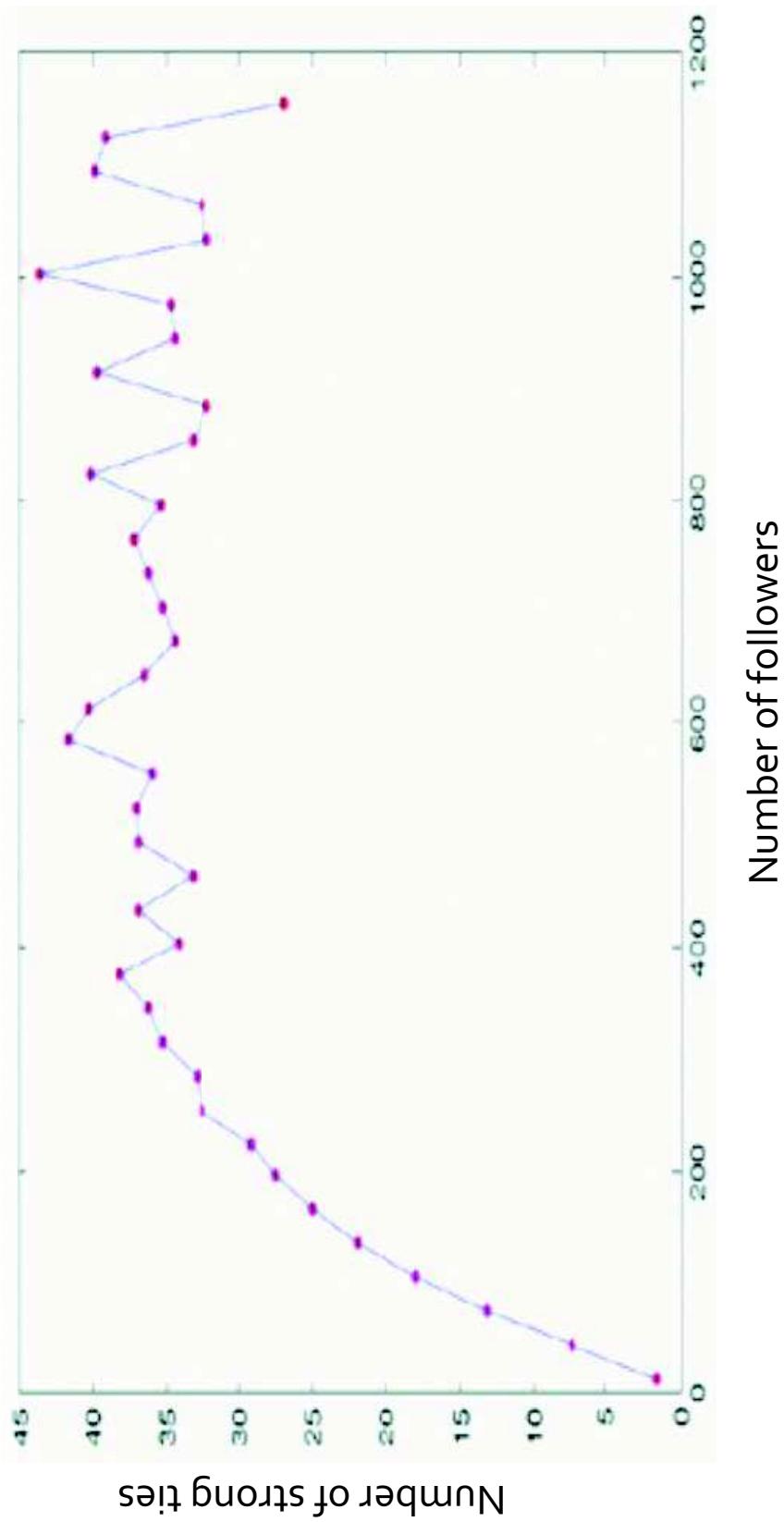
[Marlow et al. '09]

Active Network Sizes



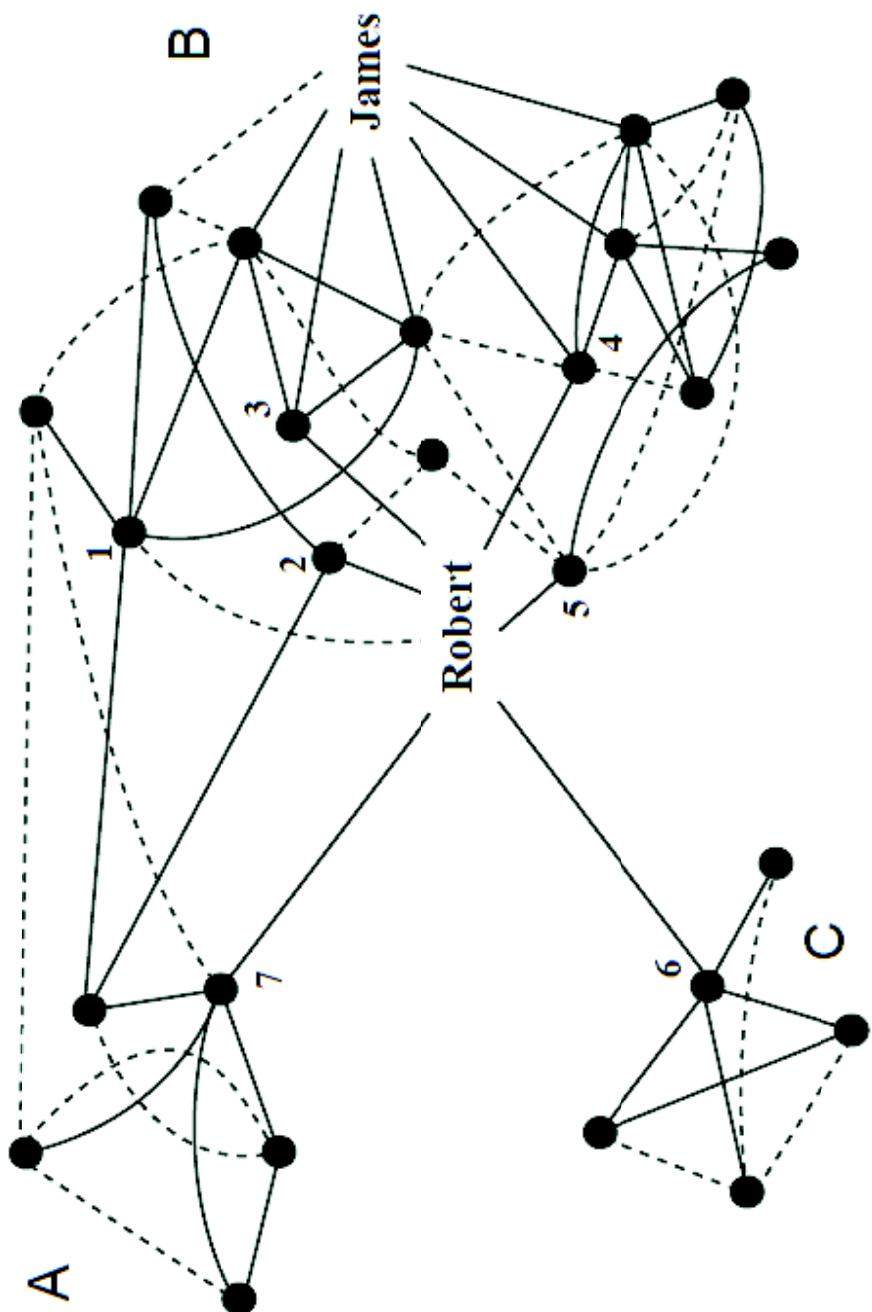
[Huberman et al. '09]

# Twitter: Strong ties vs. Followers

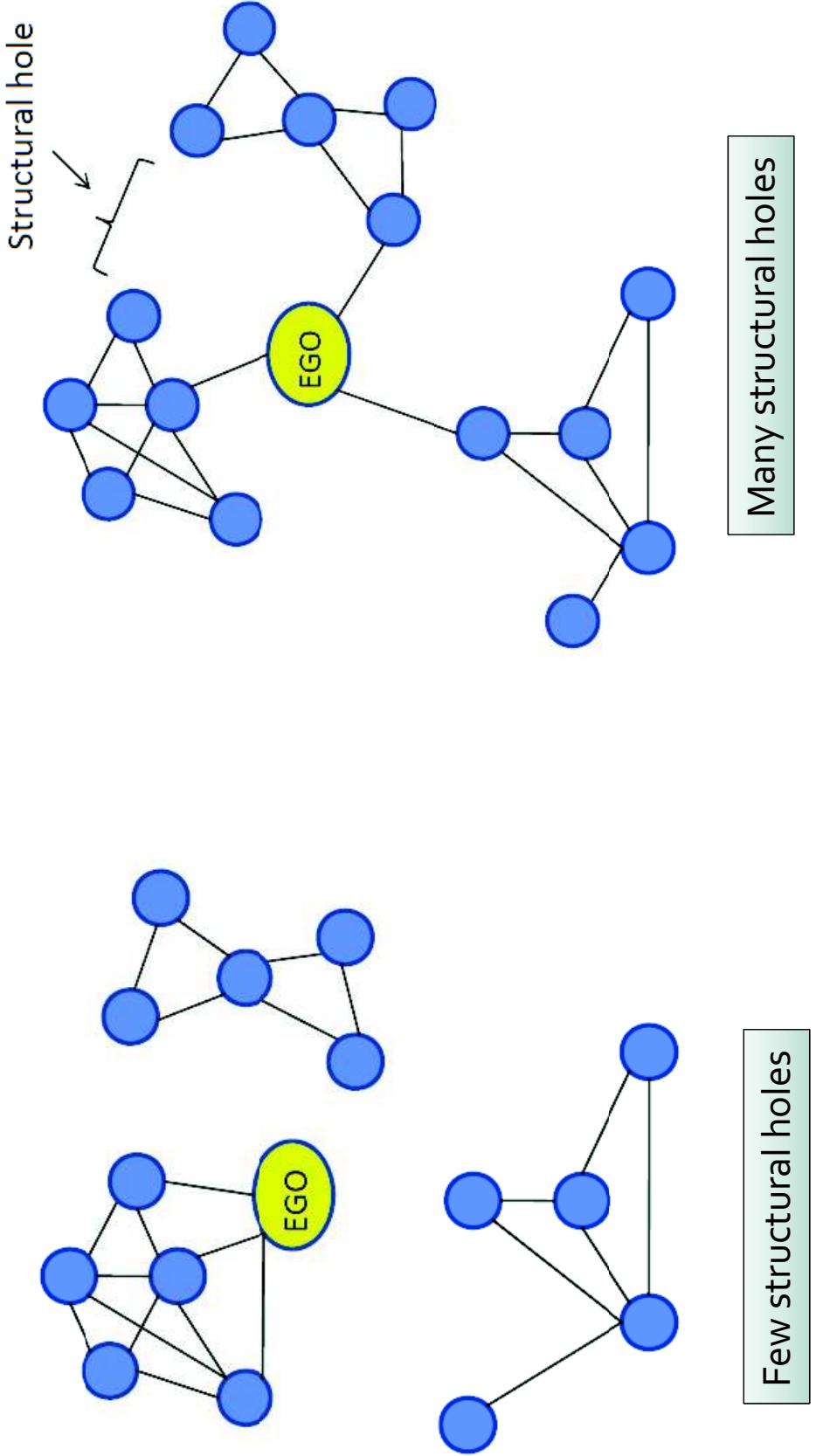


# Structural Holes

[Ron Burt]



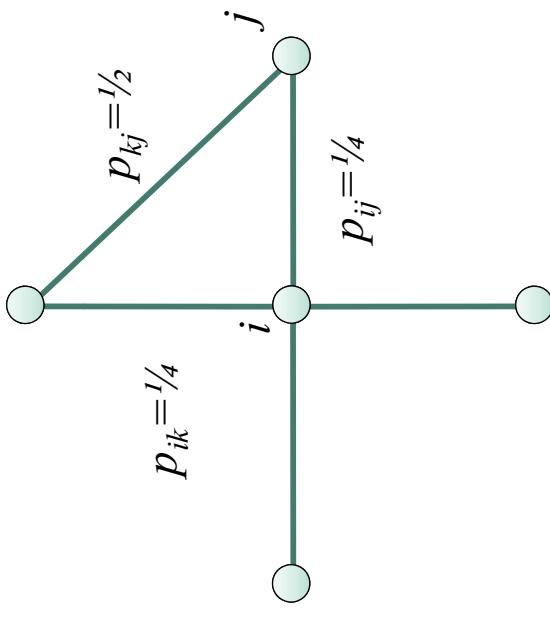
# Structural Holes



Structural Holes provide Ego with access to novel information, power, freedom

# Structural Holes: Network constraint

- The ‘network constraint’ measure [Burt]:
  - To what extent are person’s contacts redundant
    - Low: disconnected contacts
    - High: contacts that are close or strongly tied



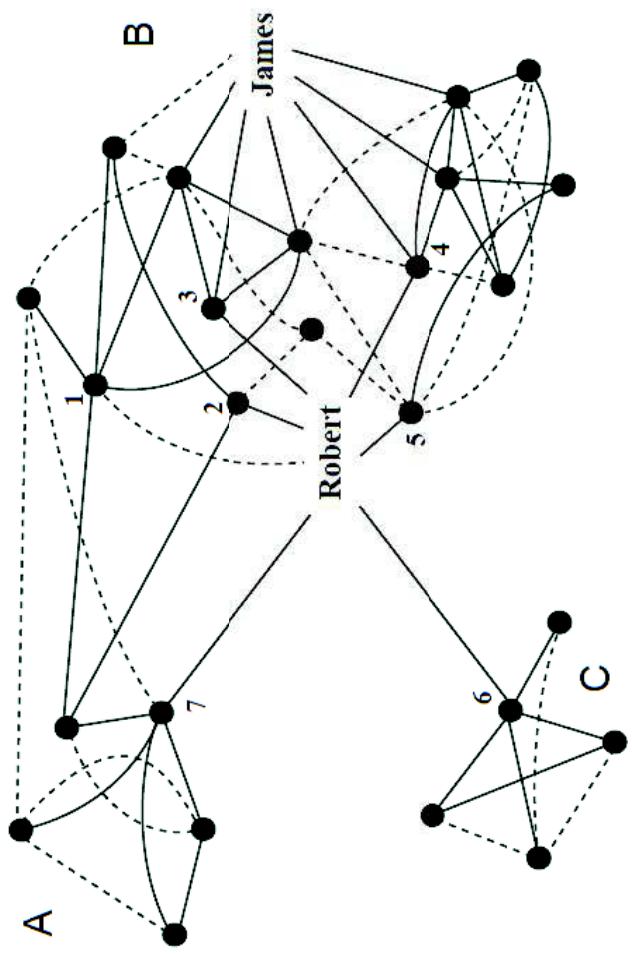
$$c_i = \sum_j \left[ p_{ij} + \sum_k (p_{ik} p_{kj}) \right]^2 :$$

$$c_i = \left[ \frac{1}{4} + \left( \frac{1}{4} \frac{1}{2} \right) \right]^2 + \left[ \frac{1}{4} + \left( \frac{1}{4} \frac{1}{2} \right) \right]^2 + \left[ \frac{1}{4} \right]^2 + \left[ \frac{1}{4} \right]^2 = \frac{13}{32}$$

$p_{ij}$  ... proportion of i's energy invested in relationship with j

# Example: Robert vs. James

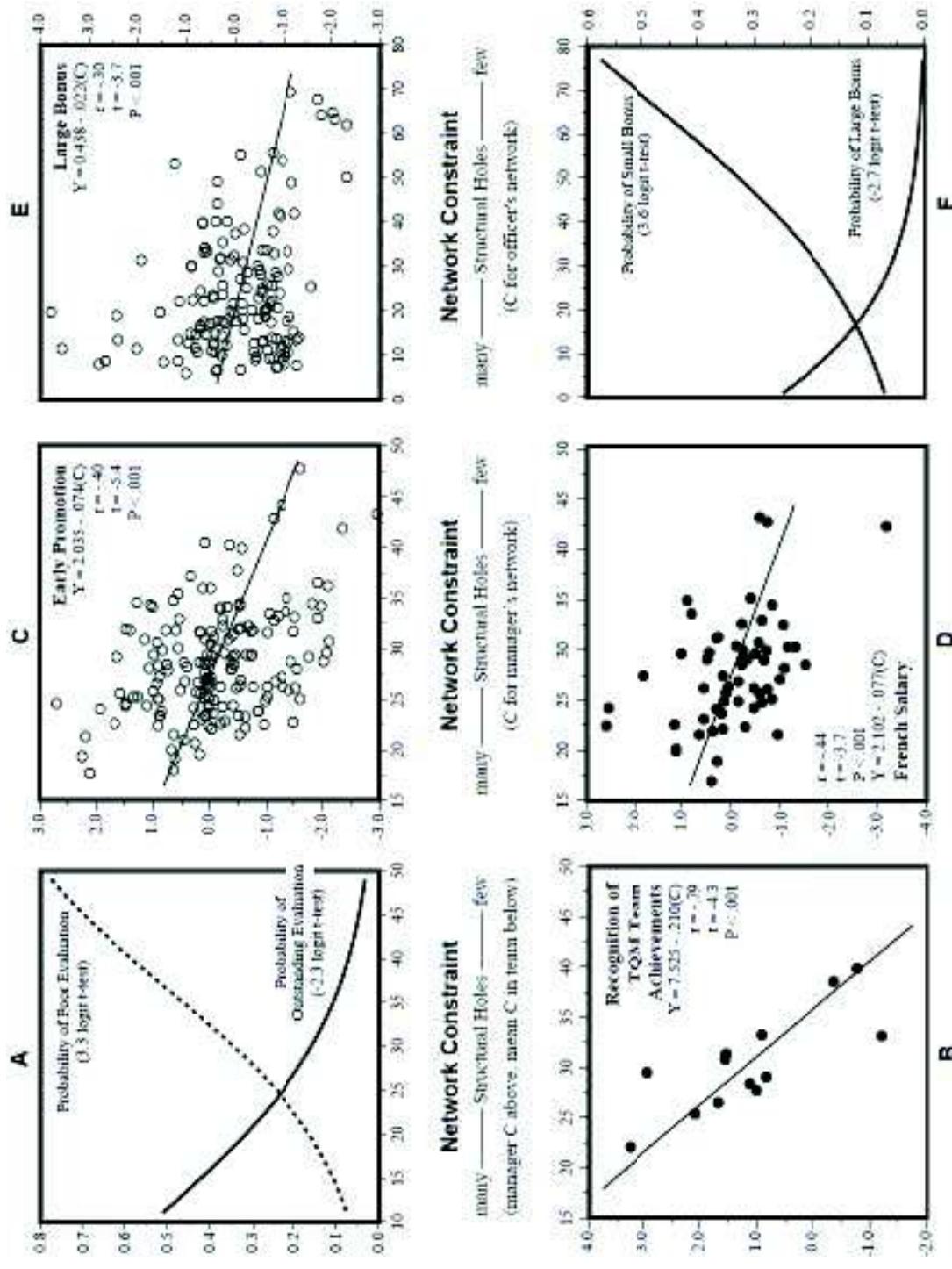
- **Constraint:** To what extent are person's contacts redundant
  - **Low:** disconnected contacts
  - **High:** contacts that are close or strongly tied



- **Network constraint:**
  - James:  $c_j = 0.309$
  - Robert:  $c_r = 0.148$

# Spanning the holes matters

[Ron Burt]

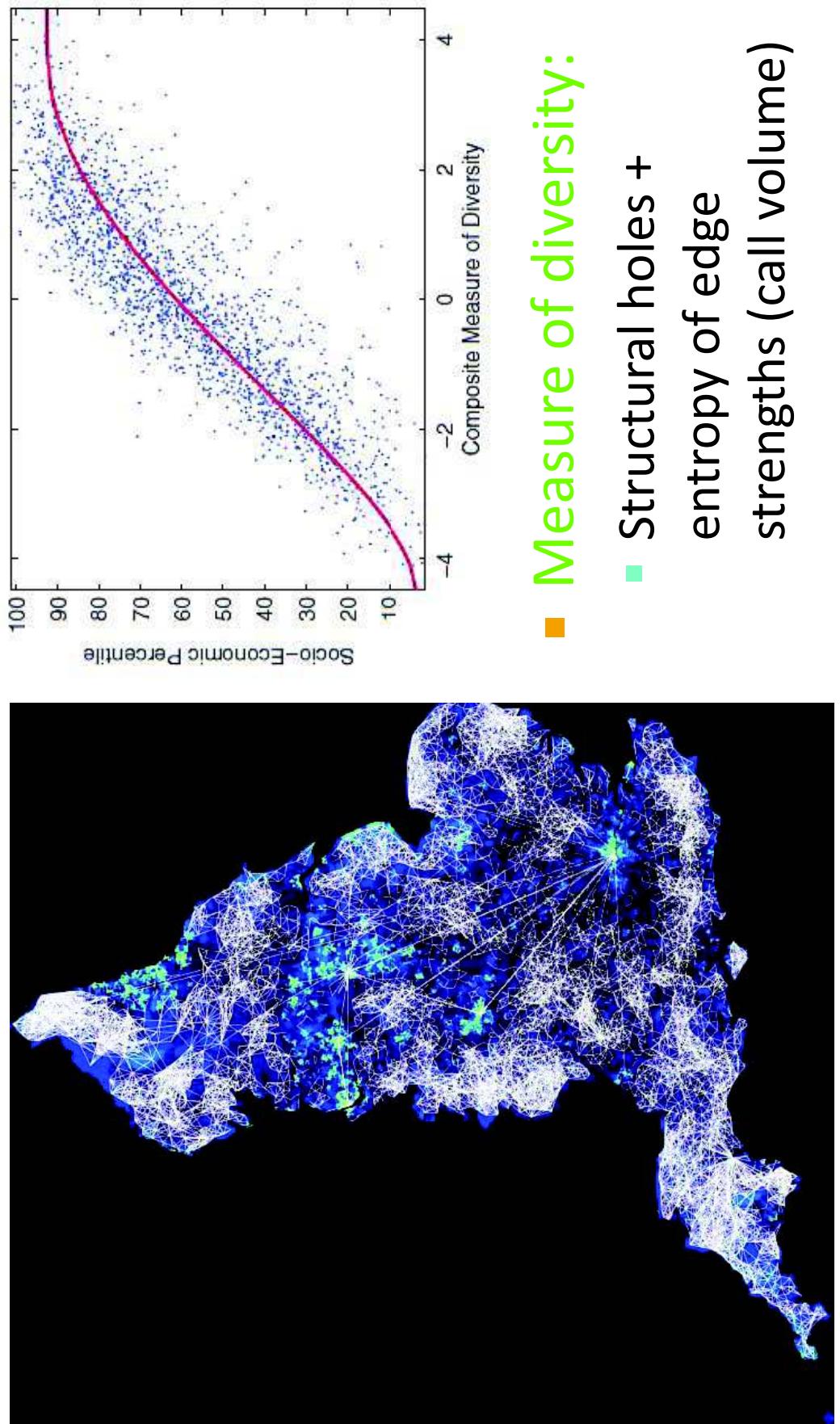


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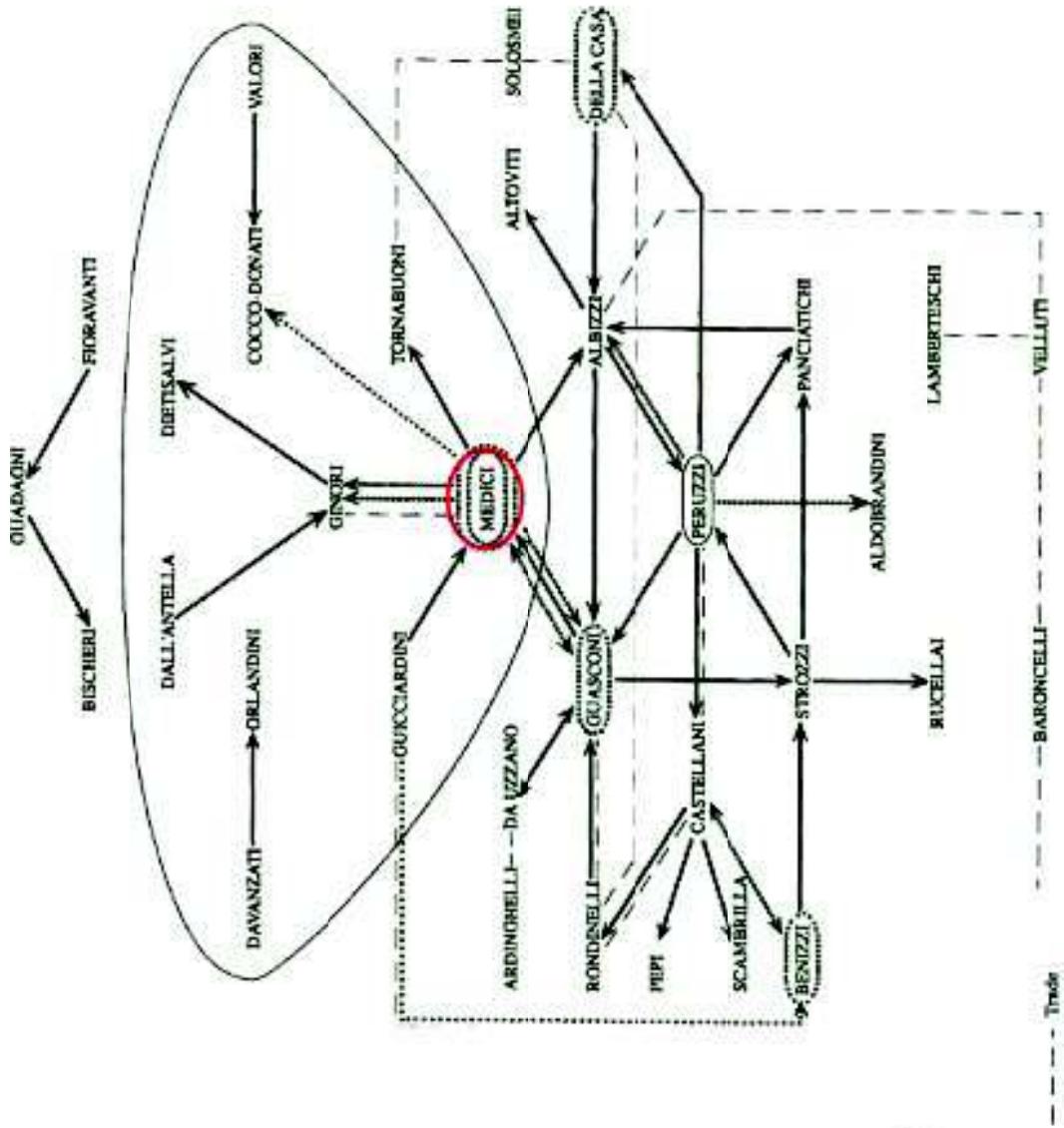
Jure Leskovec, Stanford CS224W: Social and Information Network Analysis, <http://cs224w.stanford.edu>

# Diversity & Development

[Eagle-Macy, 2010]



# Florentine families: Power



(Padgett & Ansell, 1993)

Types of Ties:

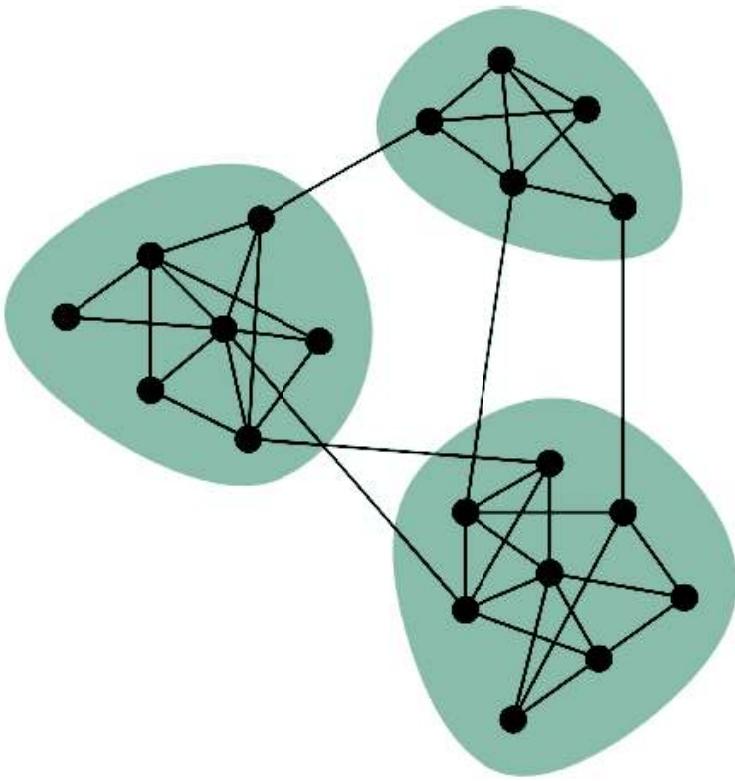
- ♂→♀ Marriage
- ↔ Partnership
- Trade
- Real Estate
- Bank Employment

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# Network communities

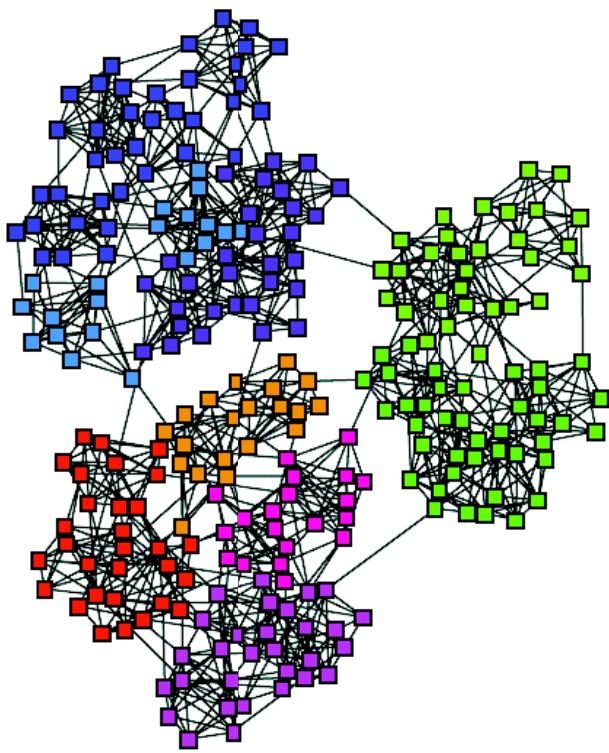
- Networks of **tightly connected groups**
- **Network communities:**
  - Sets of nodes with **lots** of connections **inside** and **few** to **outside** (the rest of the network)



Communities, clusters,  
groups, modules

# Finding network communities

- How to automatically find such densely connected groups of nodes?
- Ideally such automatically detected clusters would then correspond to real groups

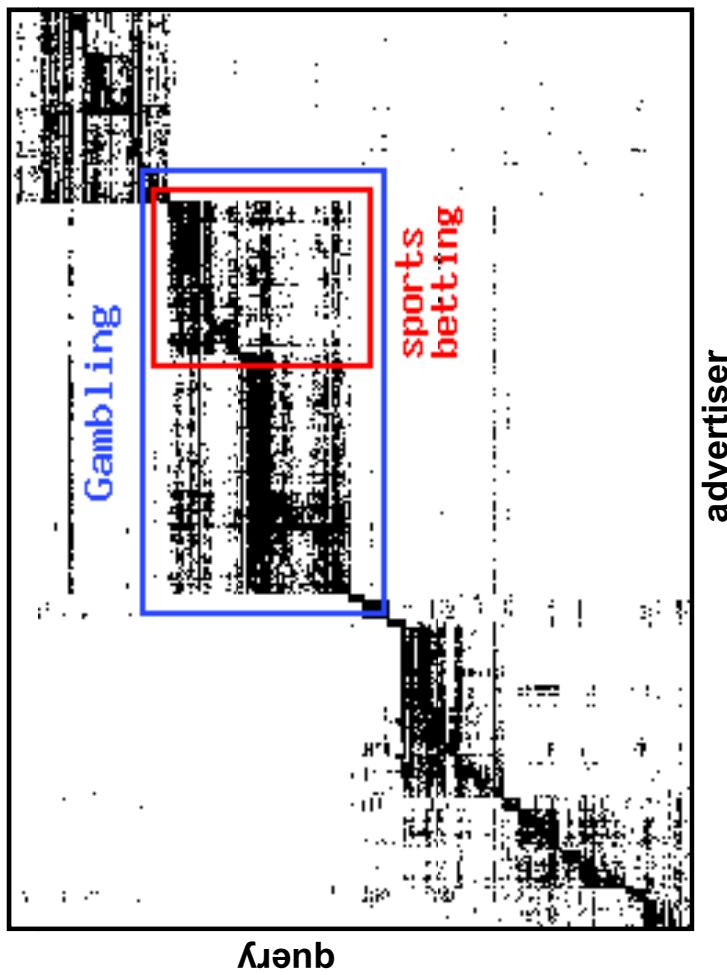


Communities, clusters, groups, modules

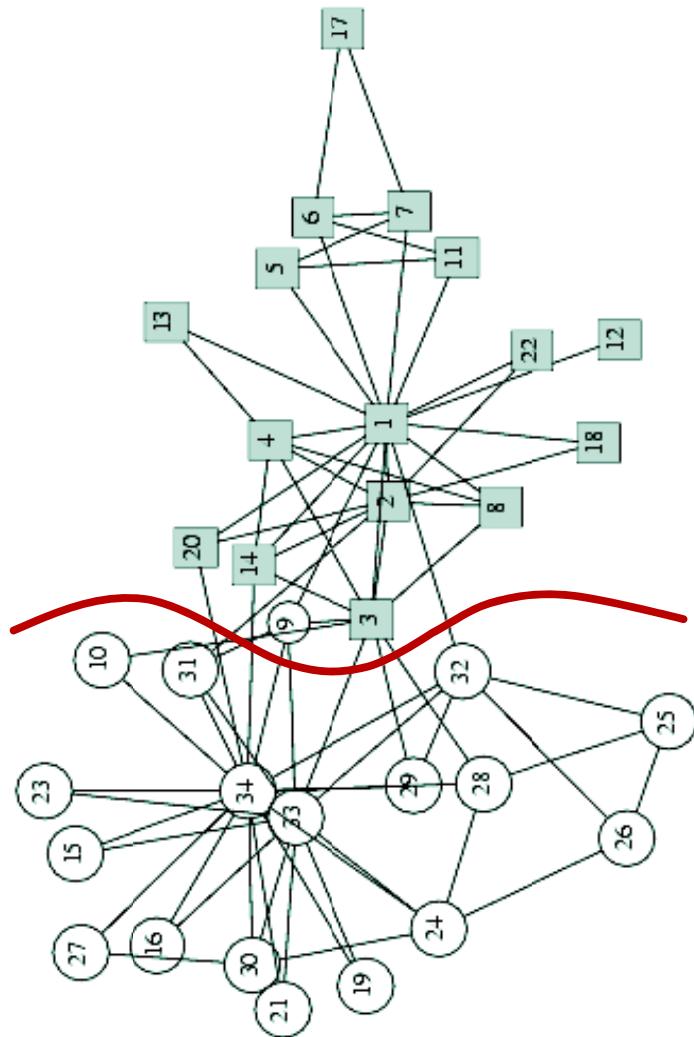
- For example:

# Micro-markets in sponsored search

Find micro-markets by partitioning the “query × advertiser” graph:



# Social Network Data



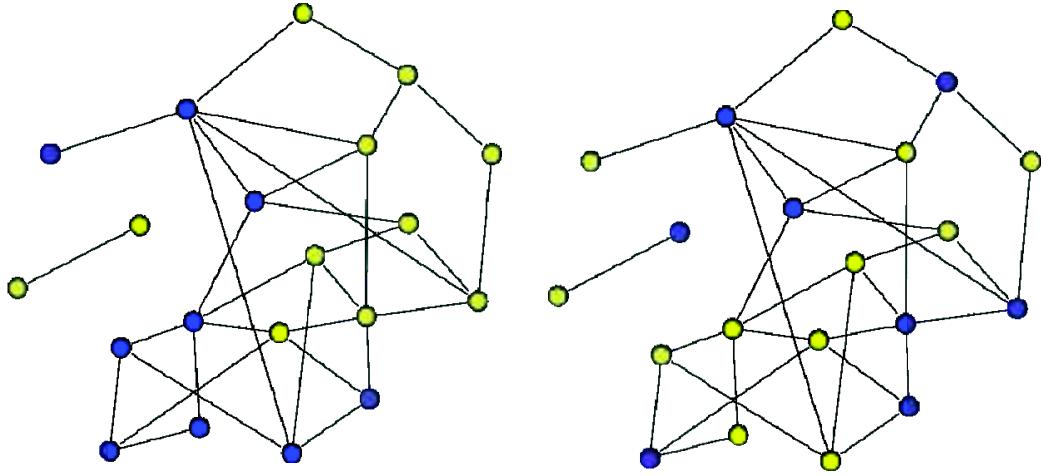
## ■ Zachary's Karate club network:

- Observe social ties and rivalries in a university karate club
- During his observation, conflicts led the group to split
- Split could be explained by a minimum cut in the network
- **Why would we expect such clusters to arise?**

# Group formation in networks

[Backstrom et al. KDD '06]

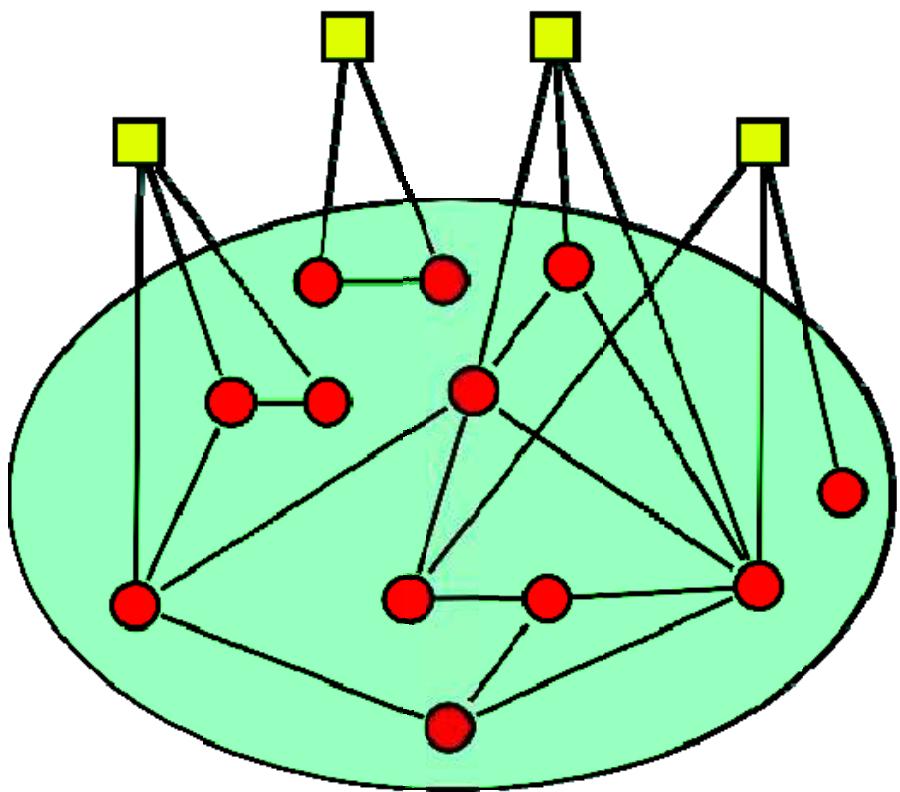
- In a social network **nodes explicitly declare group membership**:
  - Facebook groups, Publication venue
  - Can think of groups as **node colors**
- **Gives insights into social dynamics:**
  - Recruits friends? Memberships spread along edges
  - Doesn't recruit? Spread randomly
- **What factors influence a person's decision to join a group?**



# Group growth as diffusion

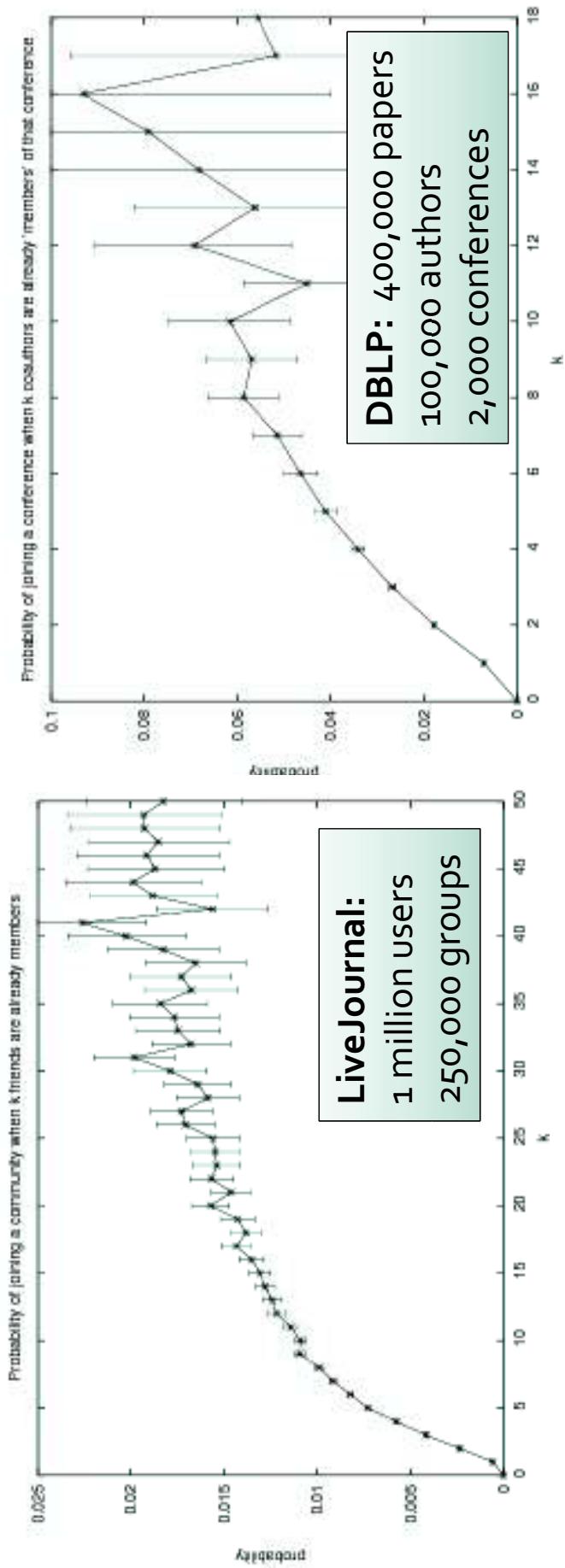
[Backstrom et al. KDD '06]

- Analogous to diffusion
- Group memberships spread over the network:
  - Red circles represent existing group members
  - Yellow squares may join
- Question:
  - How does prob. of joining a group depend on the number of friends already in the group?



# P(join) vs. # friends in the group

[Backstrom et al. KDD '06]



## ■ Diminishing returns:

- Probability of joining increases with the number of friends in the group
- But increases get smaller and smaller

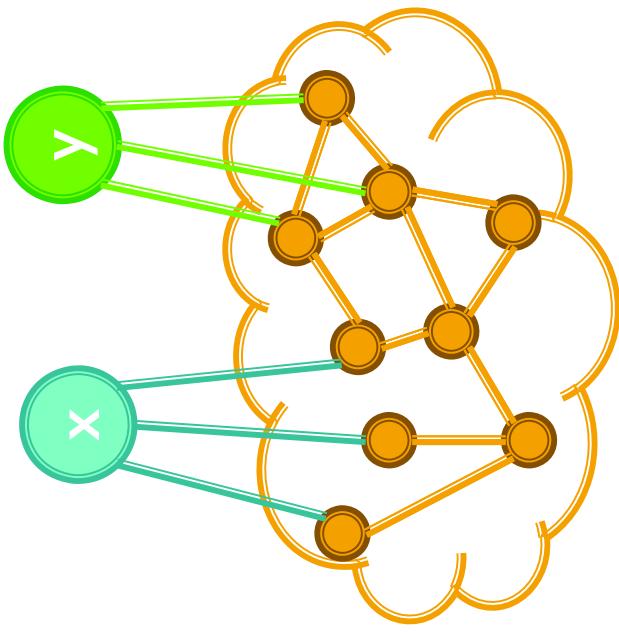
# Groups: More subtle features

[Backstrom et al. KDD '06]

## ■ Connectedness of friends:

- $x$  and  $y$  have three friends in the group
- $x$ 's friends are independent
- $y$ 's friends are all connected

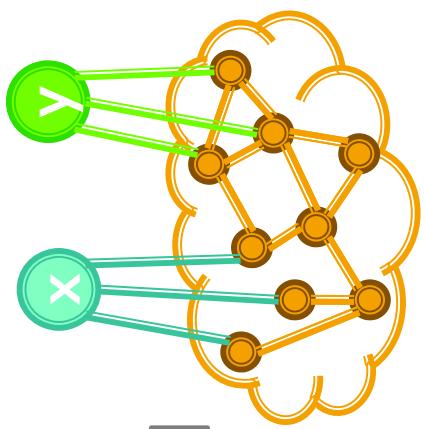
Who is more likely to join?



# Connectedness of Friends

[Backstrom et al. KDD '06]

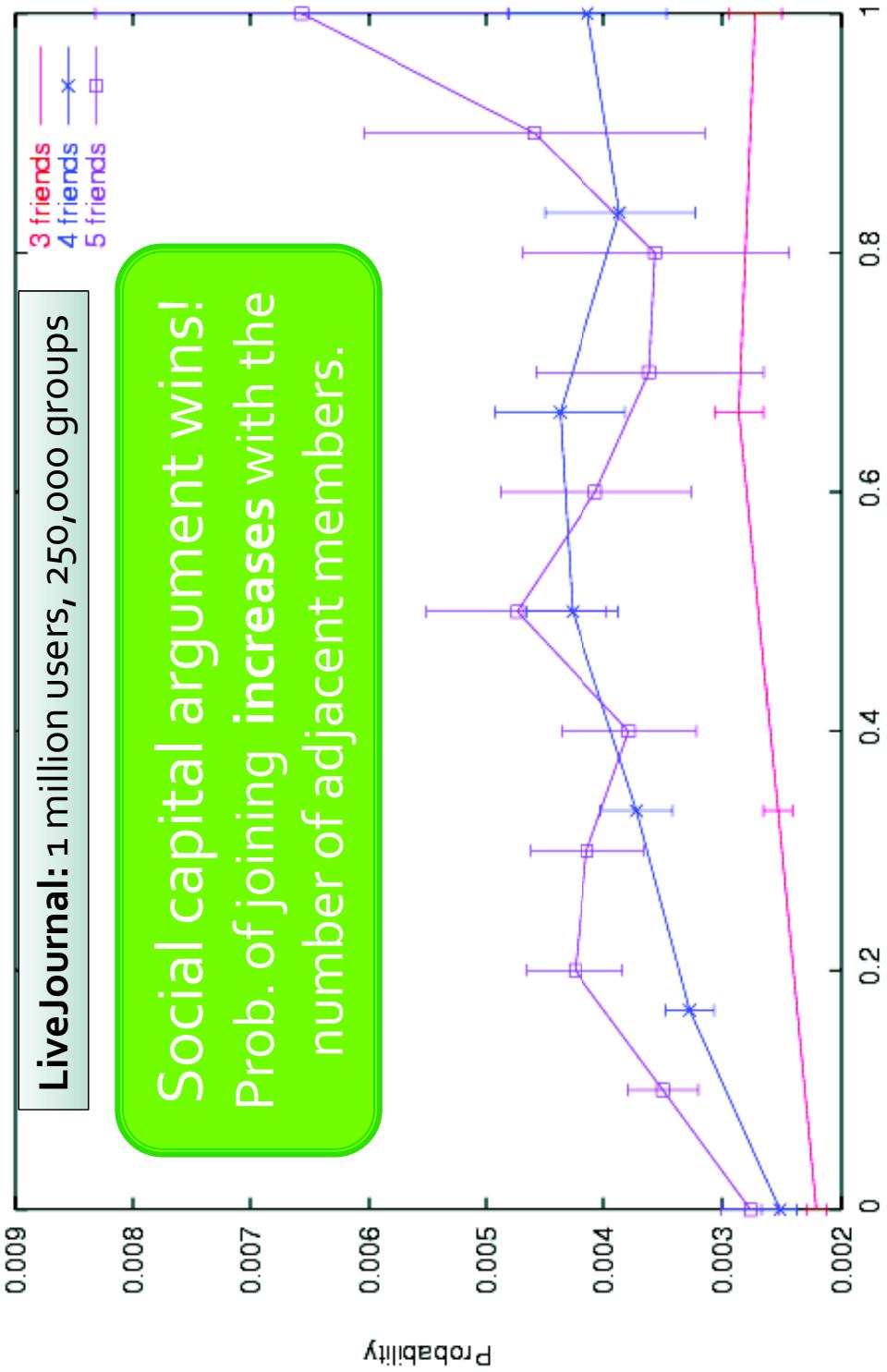
- Competing sociological theories:
  - Information argument [Granovetter '73]
  - Social capital argument [Coleman '88]
- Information argument:
  - Unconnected friends give independent support
- Social capital argument:
  - Safety/trust advantage in having friends who know each other



# Connectedness of friends

[Backstrom et al. KDD '06]

Probability of joining a community versus adjacent pairs of friends in the community



# So, this means that

- A person is more likely to join a group if
  - she has more friends who are already in the group
  - friends have more connections between themselves
- So, groups form clusters of tightly connected nodes

