

VE444: Networks

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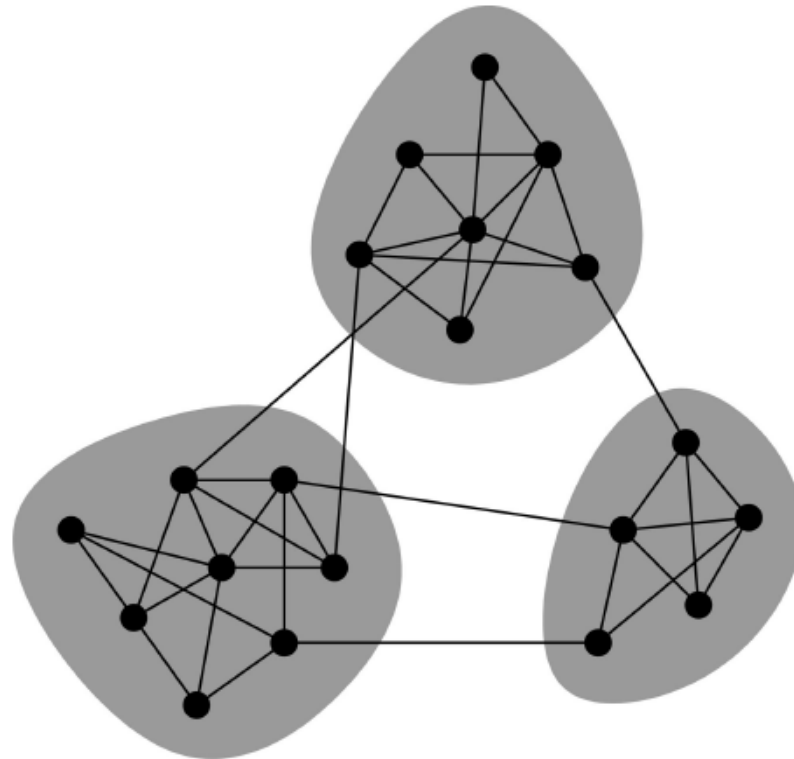
Strong and weak ties

Acknowledgment:
**This lecture's slides are modified
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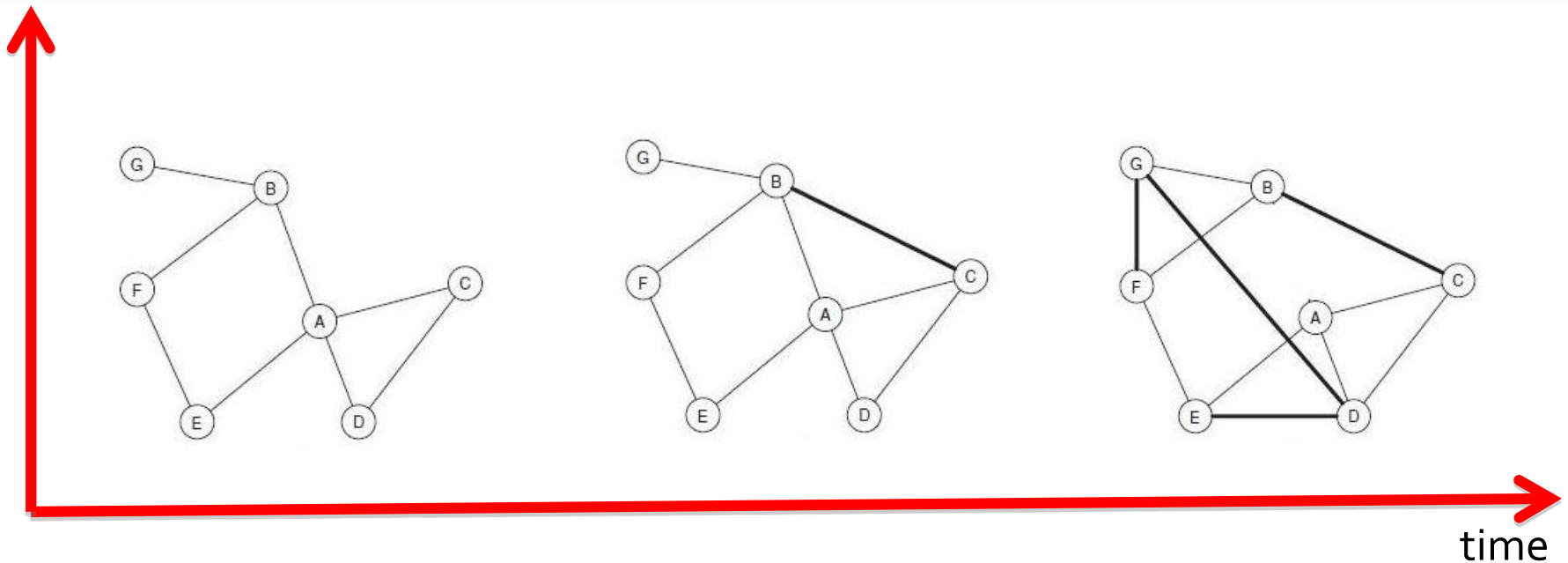
Networks & Communities

- We often think of networks “looking” like this:



- What led to such a conceptual picture?

What if we consider the evolution of the a network



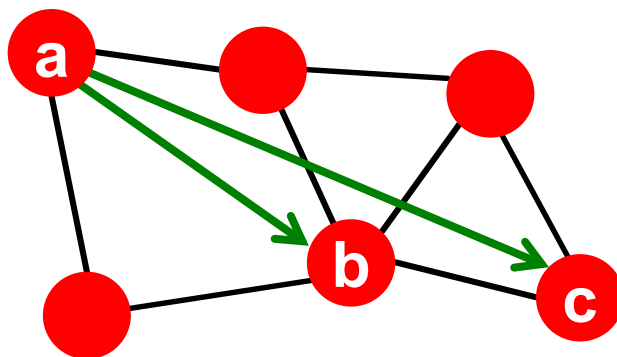
- Consider not only a snapshot, but evolution

Motivating example

- **Network could facilitate the flow of information**
- **How do 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
- **Why is it that acquaintances are most helpful?**

Granovetter's Answer

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

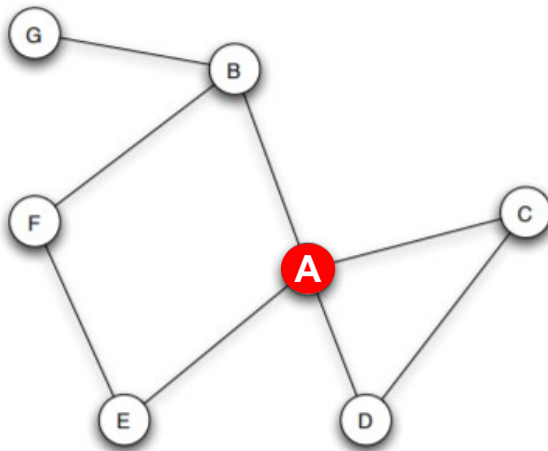


Which edge is more likely, a-b or a-c?

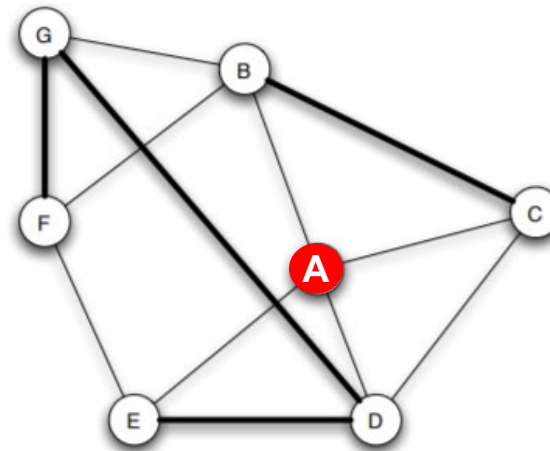
If two people in a network have a friend in common, then there is an increased likelihood they will become friends themselves.

Triadic Closure

- Motivation of the clustering coefficient



(a) Before new edges form.



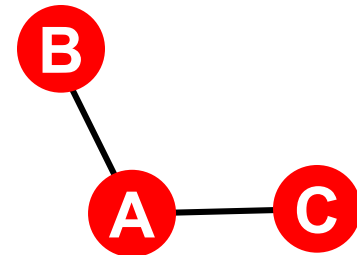
(b) After new edges form.

- **Triadic closure** = **High clustering coefficient**

Reasons for Triadic Closure

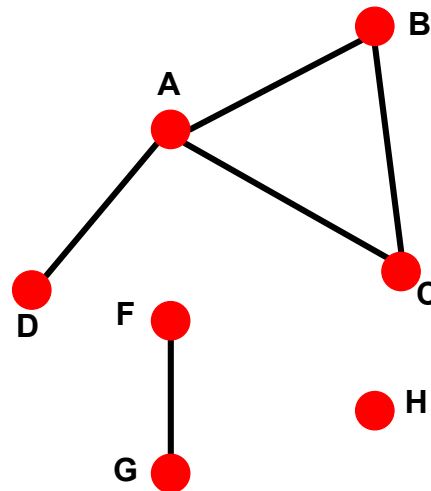
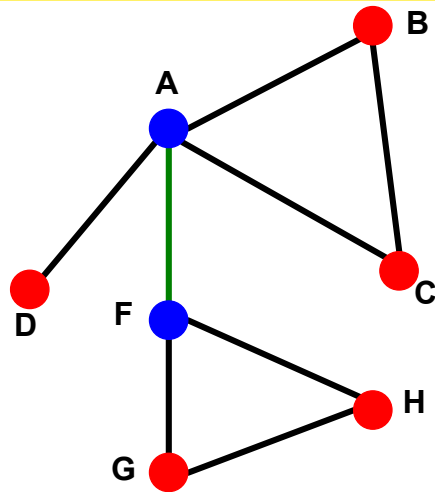
Reasons for triadic closure:

- 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
 - (since 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

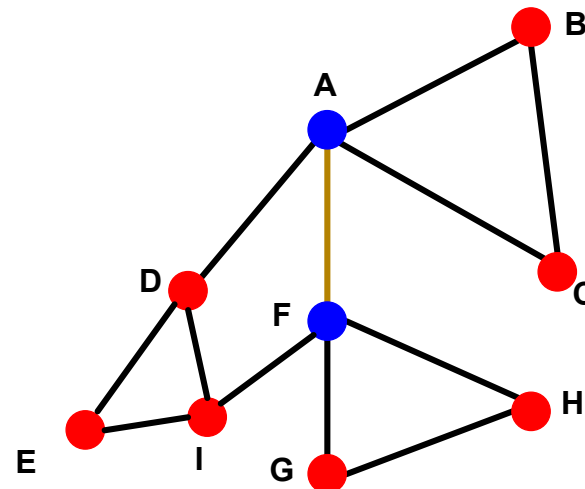
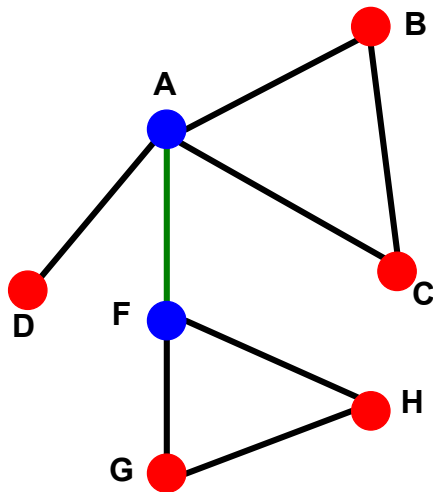
- How does triadic closure relates to the Granovetter's study?
- **Bridge** describes an edge's role where deleting this edge would cause its corresponding vertices falling in different components;



Articulation node: If we erase the **node**, the graph becomes disconnected

Bridges and local bridges

- An edge is a **local bridge** if its **end points** have **no friends in common**
 - (Alternatively, distance perspective?)
 - **Span** of a local bridge: the **distance its endpoints** would be from each other if the edge were deleted

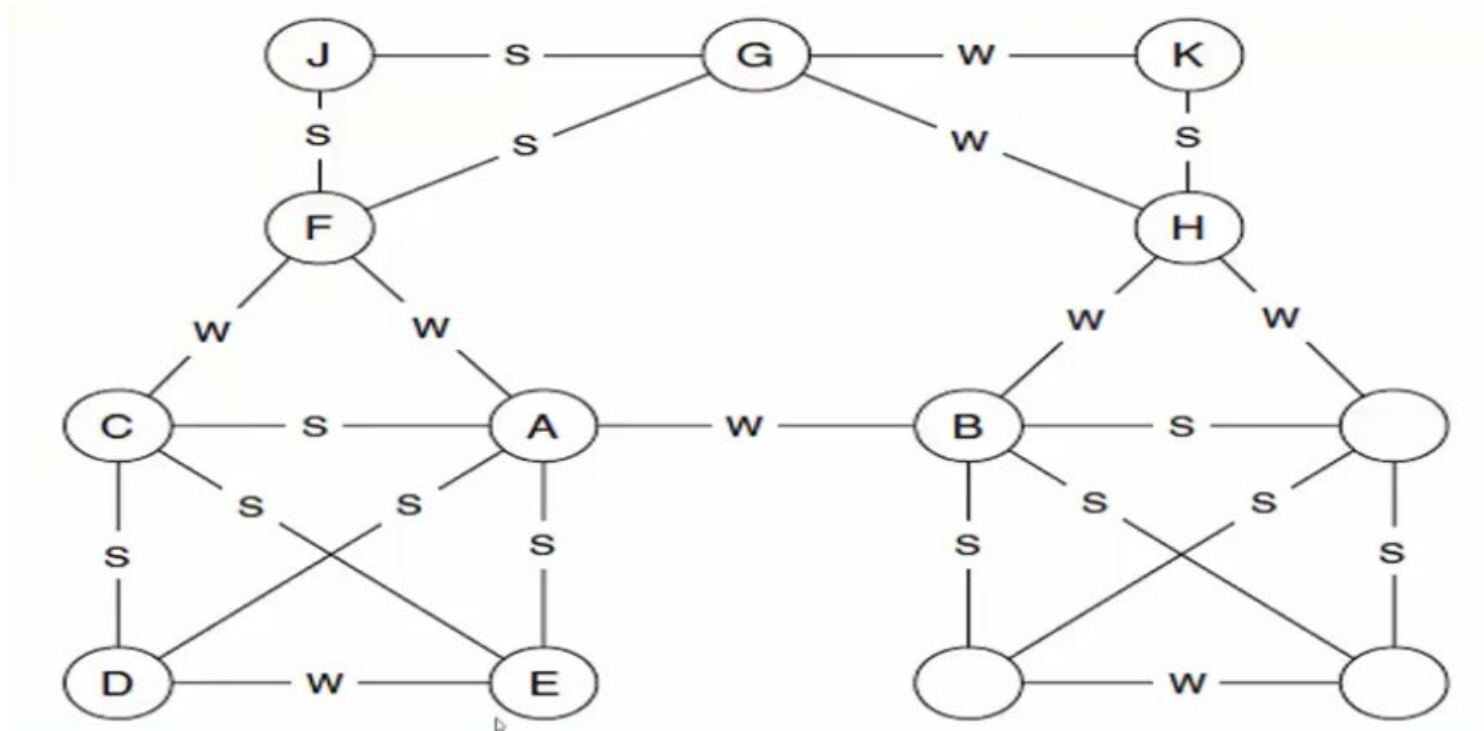


Strong and weak ties

- If we can categorize all links in the social networks as **strong ties** and **weak ties**
- Relating to the triadic closure, we make the following assumption:
 - If a node A has edges to nodes B and C, then the B-C edge is especially likely to form if A's edges to B and C are both strong ties.
 - More formally, as Granovetter suggested:
- A node A violates the **Strong Triadic Closure property** if it has strong ties to two other nodes B and C, and there is no edge at all (either a strong and weak tie) between B and C.

Strong triadic closure property

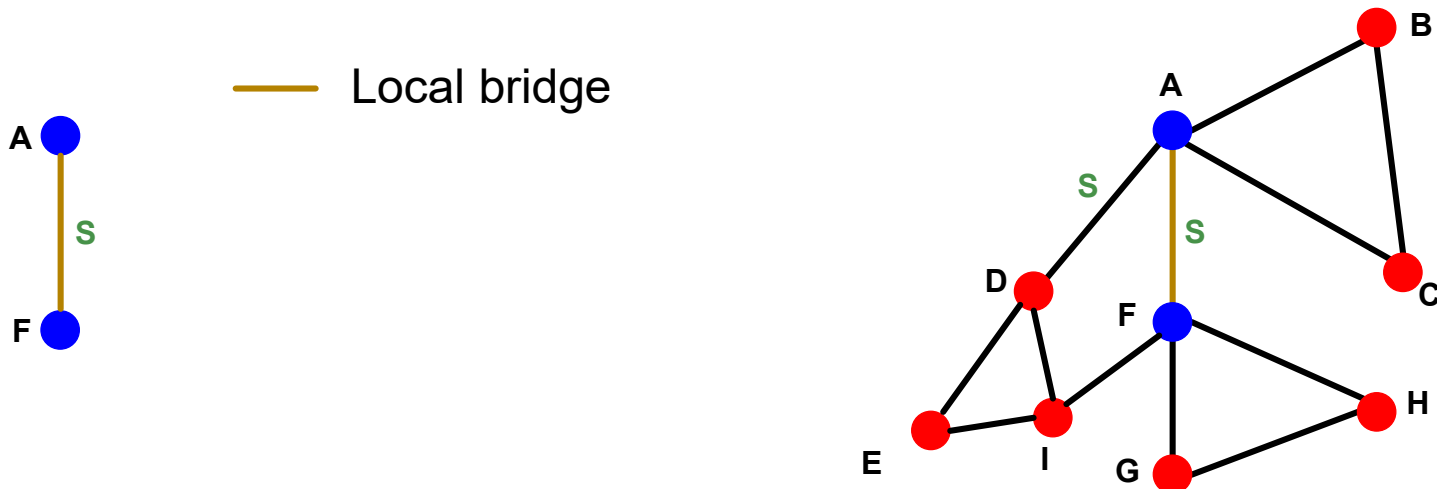
- Which node violates the strong triadic closure property?



Structural level meets interpersonal level

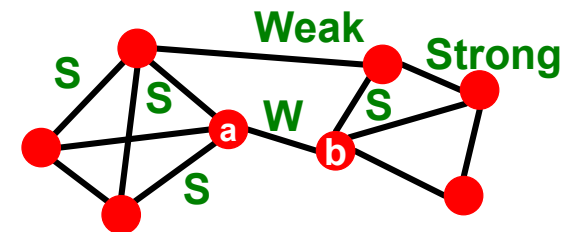
- **Bridge**: global structural notation
- **Weak/Strong**: local interpersonal distinction
- How they link to each other? Granovetter's Theorem

if a node A in a network satisfies the STC property and is involved in at least two strong ties, then any local bridge it is involved in must be a weak tie



Granovetter's Explanation

- Granovetter makes a connection between the social and structural role of an edge
- **First point: Structure**
 - Structurally embedded edges are also socially strong
 - Long-range edges spanning different parts of the network are socially weak
- **Second point: Information**
 - Long-range edges allow you to gather information from different parts of the network and get a job
 - Structurally embedded edges are heavily redundant in terms of information access



From qualitative statement to quantitative statement

- For many years 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 EU country's population
 - **Edge weight:** # phone calls

Edge Strength in Real Data

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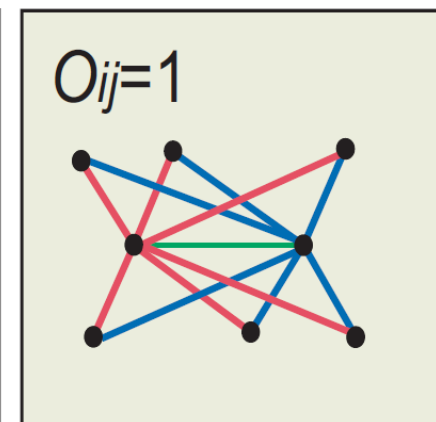
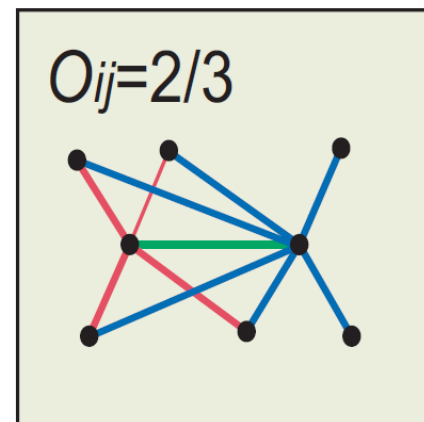
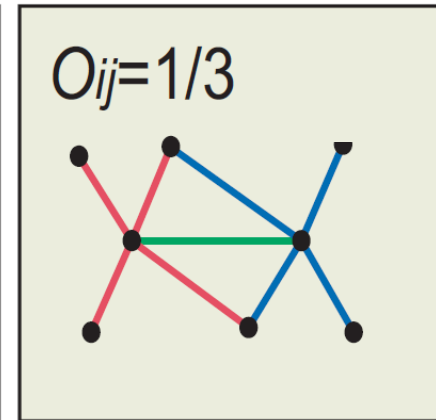
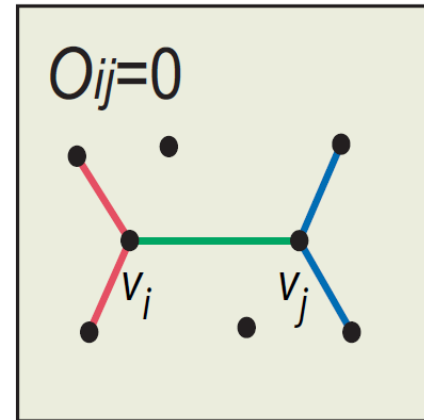
Edge Overlap

- **Edge overlap:**

$$O_{ij} = \frac{|(N(i) \cap N(j)) \setminus \{i, j\}|}{|(N(i) \cup N(j)) \setminus \{i, j\}|}$$

- $N(i)$... the set of neighbors of node i

- **Note: Overlap = 0**
when an edge is a **local bridge**



Phones: Edge Overlap vs. Strength

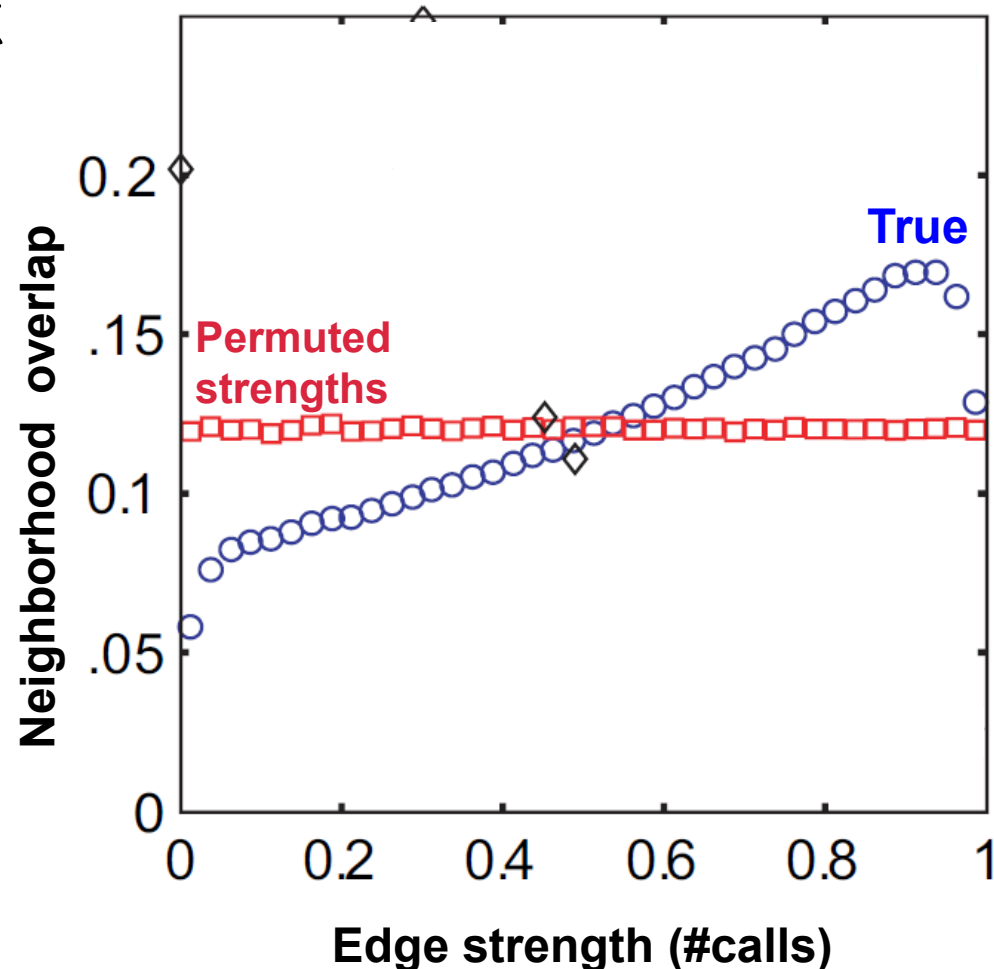
- Cell-phone network

- **Observation:**

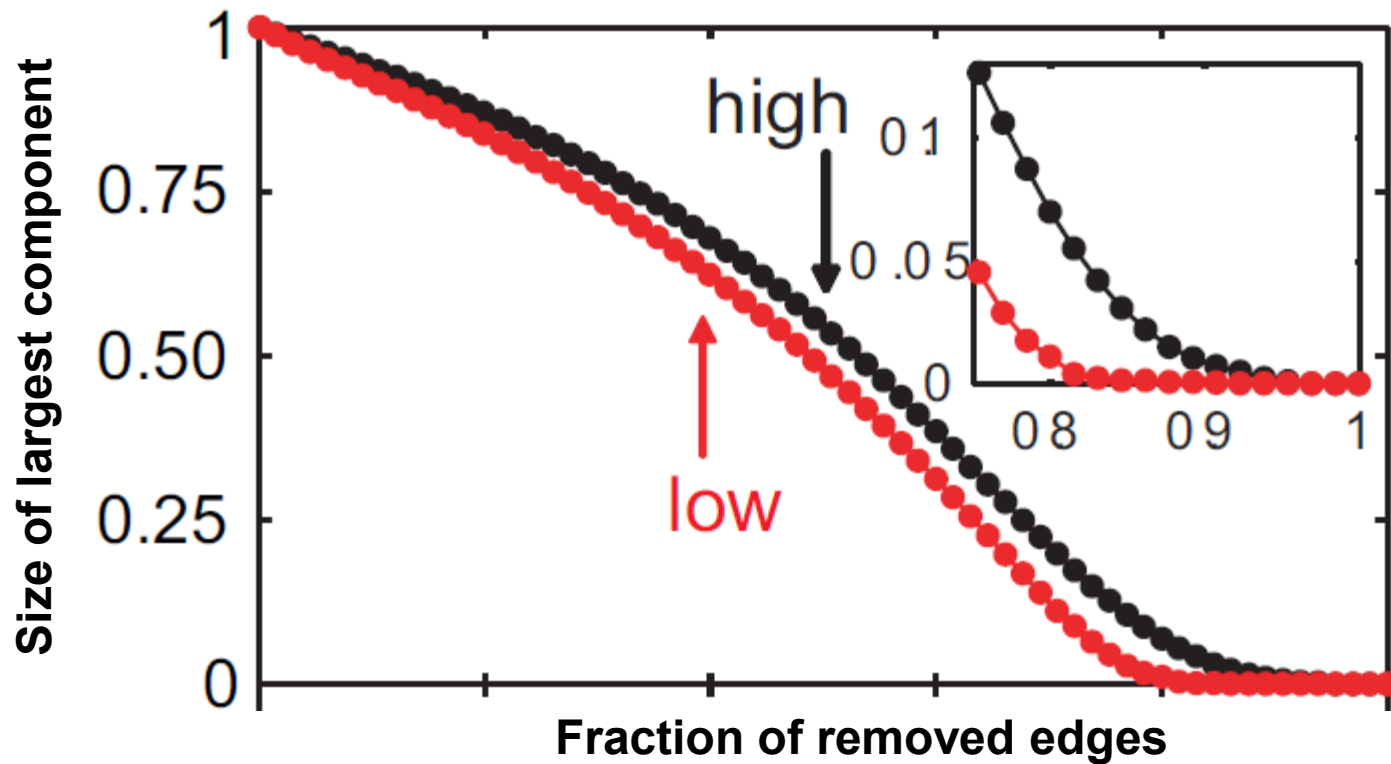
- Highly used links have high overlap!

- **Legend:**

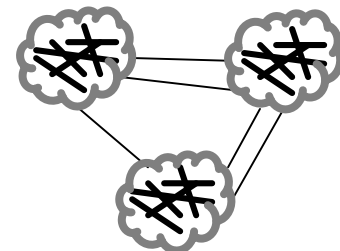
- **True:** The data
- **Permuted strengths:** Keep the network structure but randomly reassign edge strengths



Edge Removal by Strength



Low
disconnects
the network
sooner

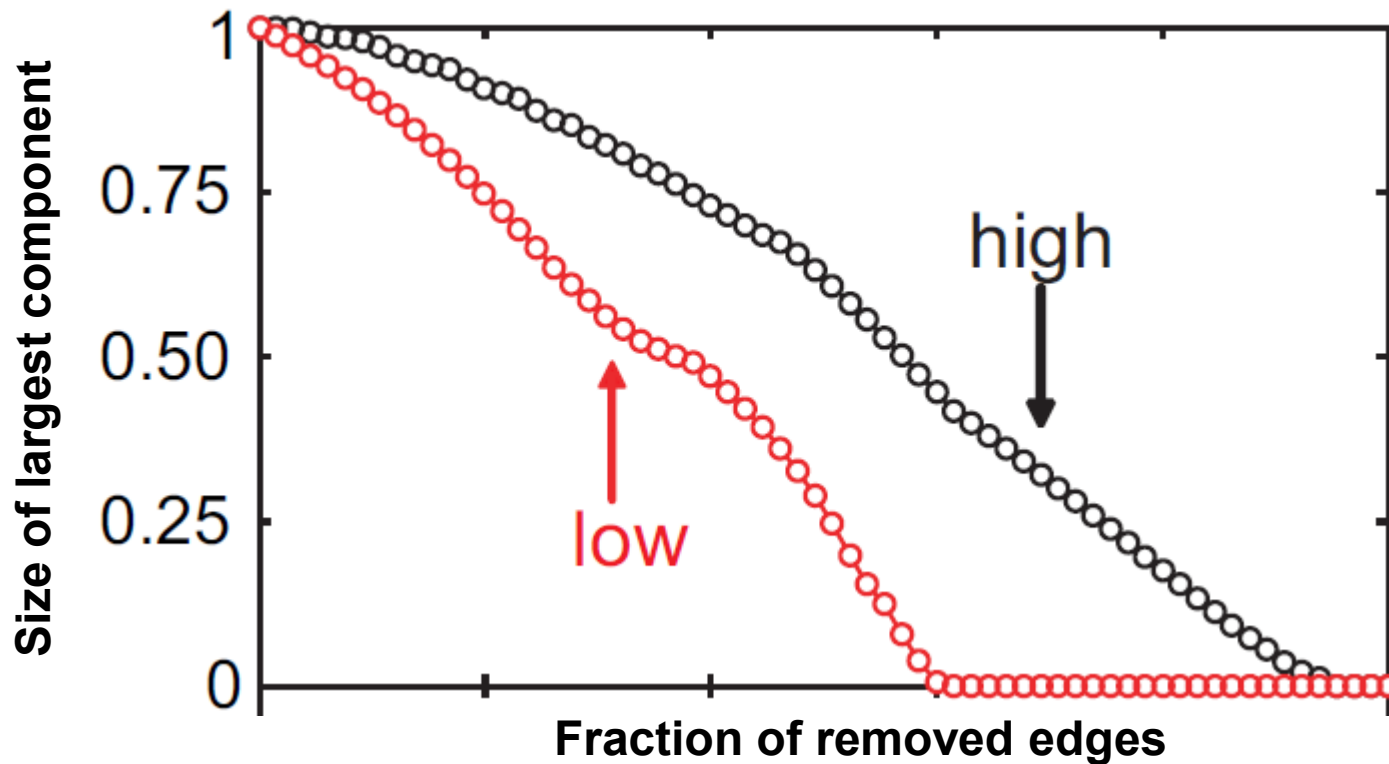


Conceptual picture
of network structure

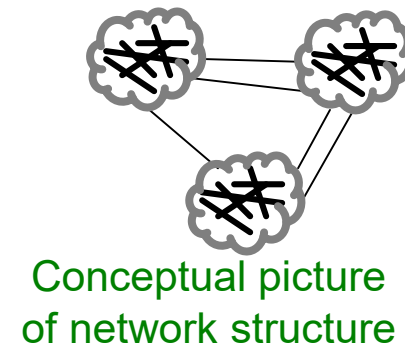
Removing edges based on **strength (#calls)**

- Low to high
- High to low

Link Removal by Overlap



Low
disconnects
the network
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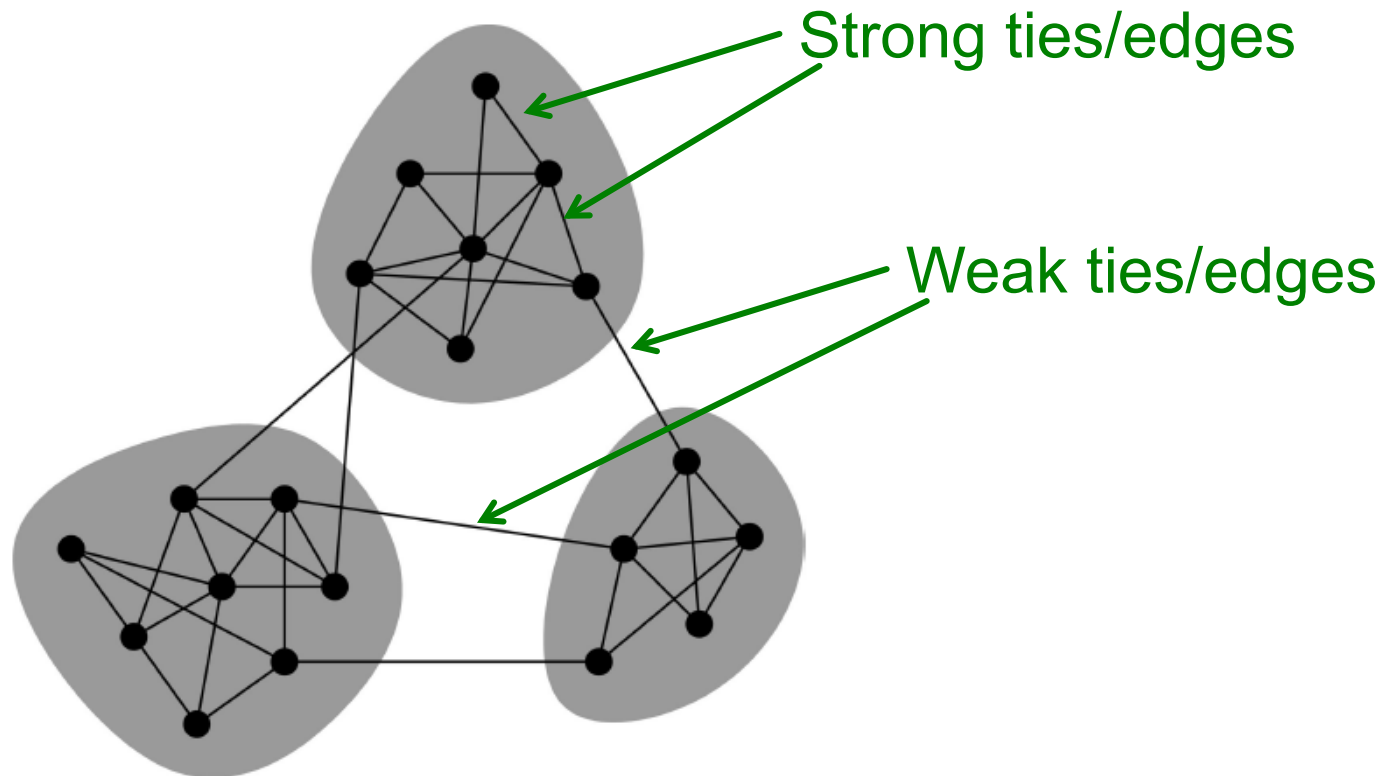


Removing edges based on **edge overlap**

- Low to high
- High to low

Conceptual Picture of Networks

- Granovetter's theory leads to the following conceptual picture of networks



Social capital

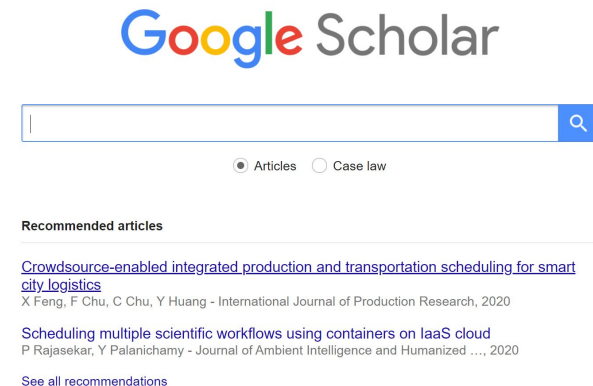
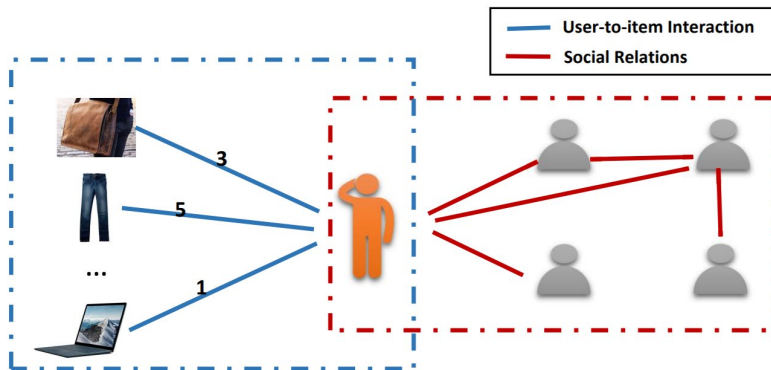
- We studied properties in graph theory, and sociology
- We saw different properties can be linked together, i.e., from one property to another property
- We studied from social science problems → abstract into graph and → study its properties (with new definitions) → validation from big data

A summary: research process of network analysis

- We studied properties in graph theory, and sociology
- We saw different properties can be linked together, i.e., from one property to another property
- We studied from social science problems → abstract into graph → study its properties (with new definitions) → validation from big data

Weak ties for recommendations

- **Social recommendations:** leverage social network information to help mitigate the “cold start” problem



Homophily

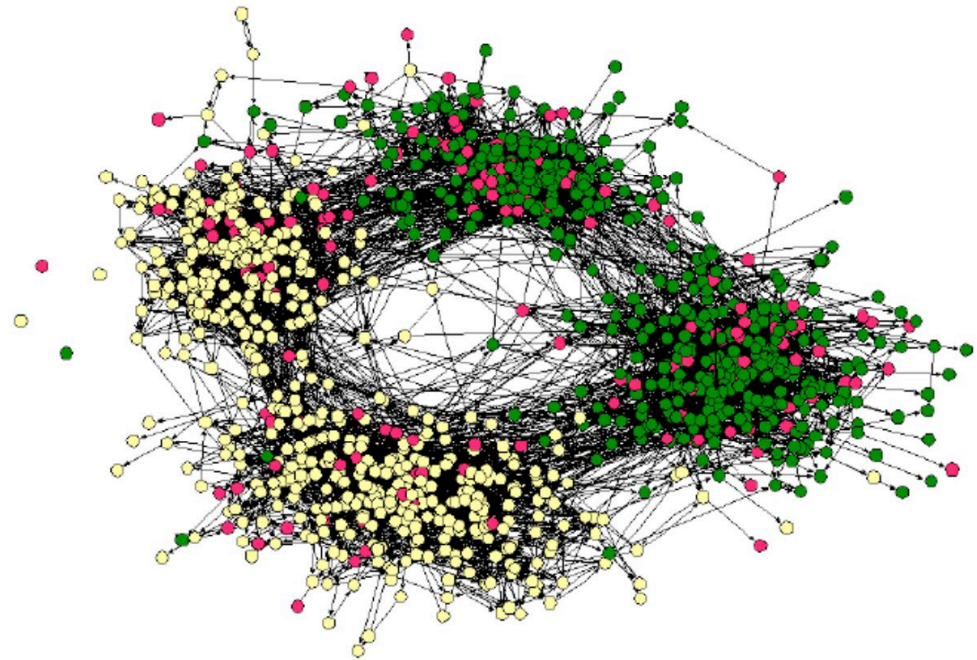
Homophily

- **Homophily**: the tendency of individuals to associate and bond **with similar others**
 - *“Birds of a feather flock together”*
 - It has been observed in a vast array of network studies, based on a variety of attributes (e.g., age, gender, organizational role, etc.)
 - **Example**: people who like the same music genre are **more likely to establish a social connection** (meeting at concerts, interacting in music forums, etc.)

Correlations Exists in Networks

Example:

- Real social network
 - Nodes = people
 - Edges = friendship
 - Node color = race
- People are segregated by race due to homophily



(Easley and Kleinberg, 2010)