

# Your friends are better than you: Friendship Paradox and its social consequences

**Kristina Lerman**

**USC Information Sciences Institute**

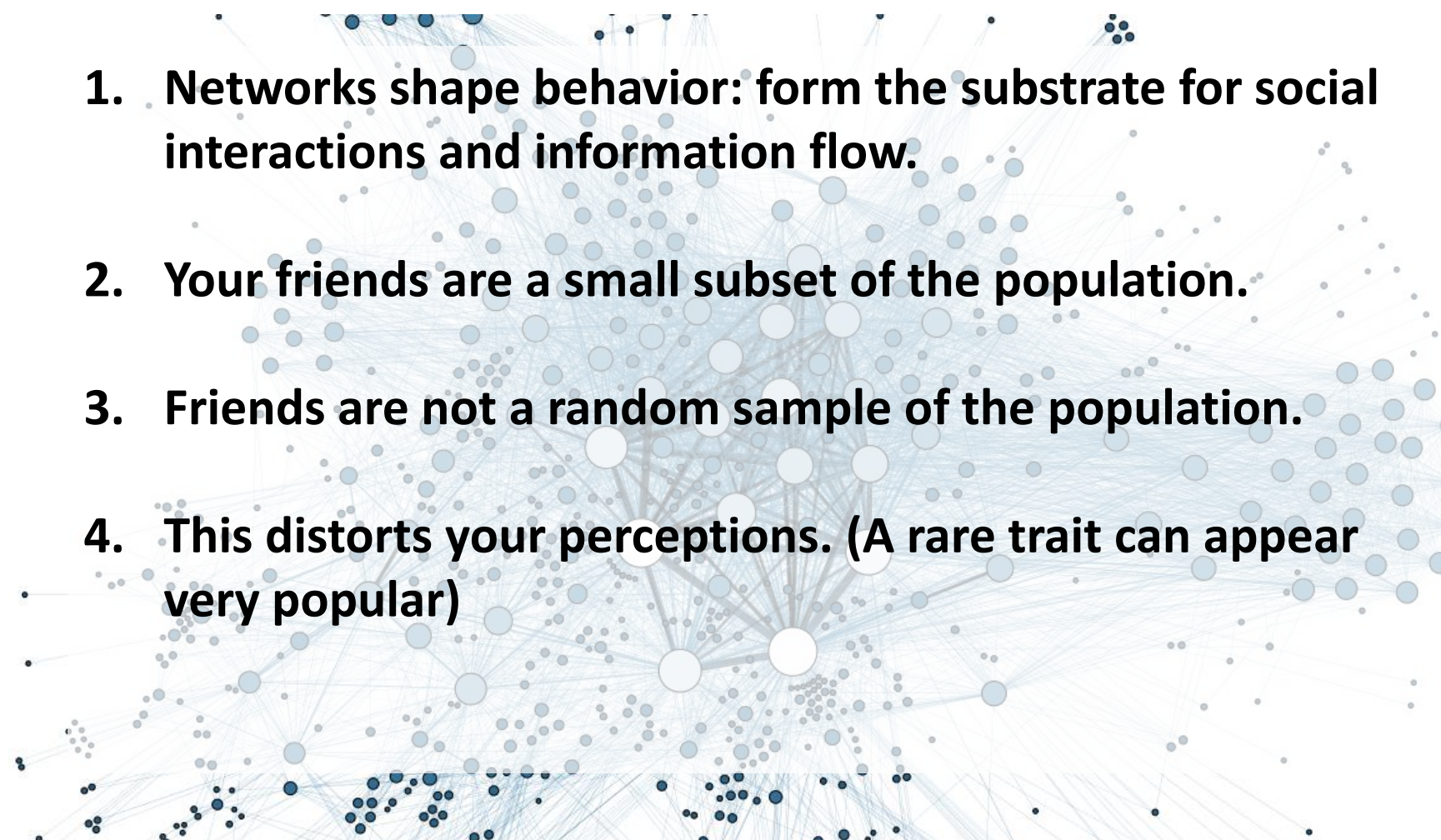
<http://www.isi.edu/~lerman>



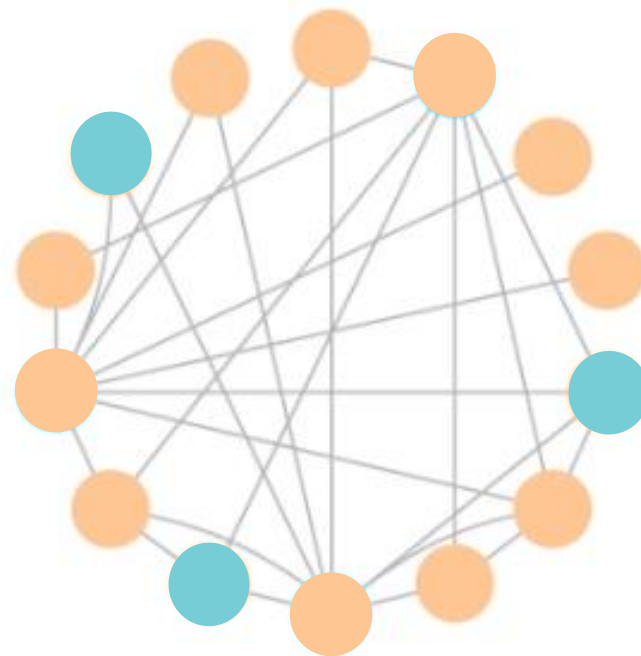
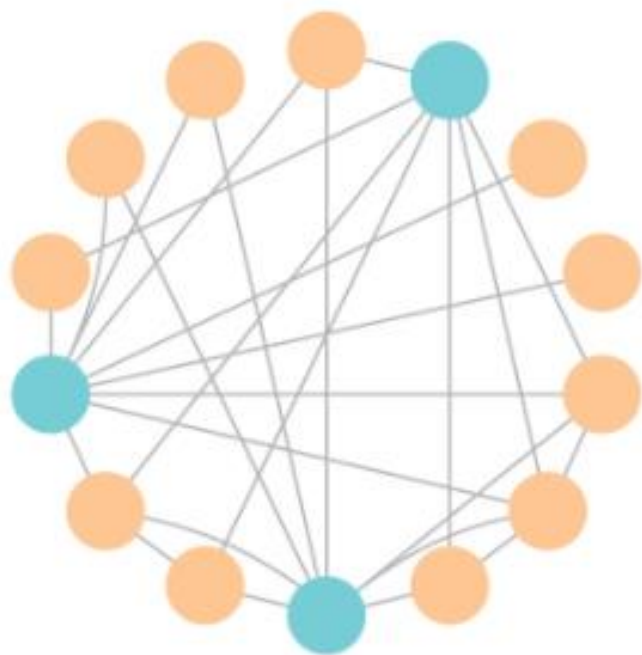
# Networks shape behavior and perceptions



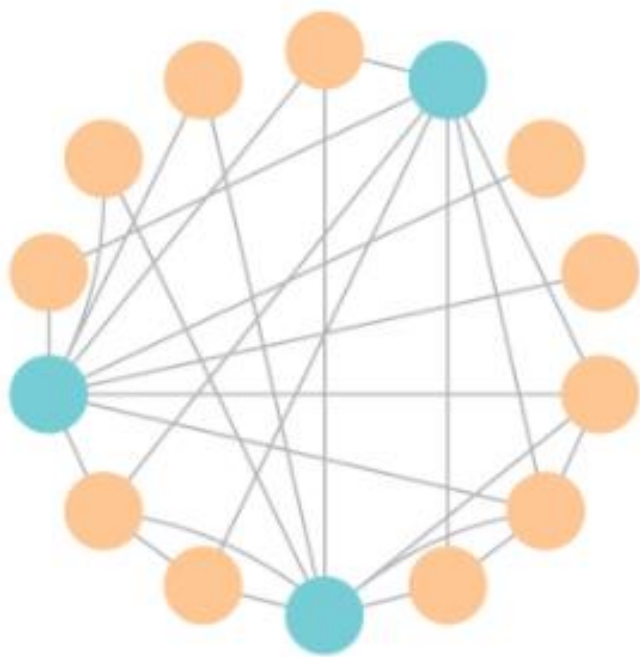
# Networks shape behavior and perceptions

- 
1. **Networks shape behavior: form the substrate for social interactions and information flow.**
  2. **Your friends are a small subset of the population.**
  3. **Friends are not a random sample of the population.**
  4. **This distorts your perceptions. (A rare trait can appear very popular)**

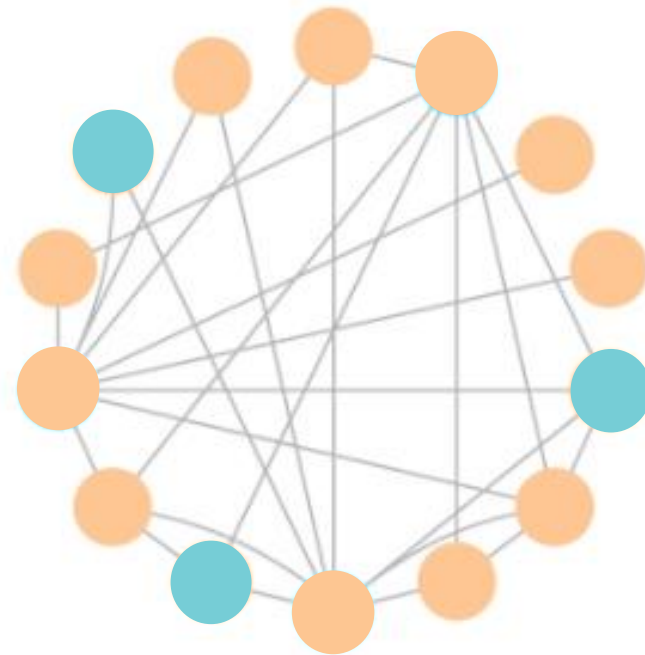
# What color is more popular?



# What color is more popular?



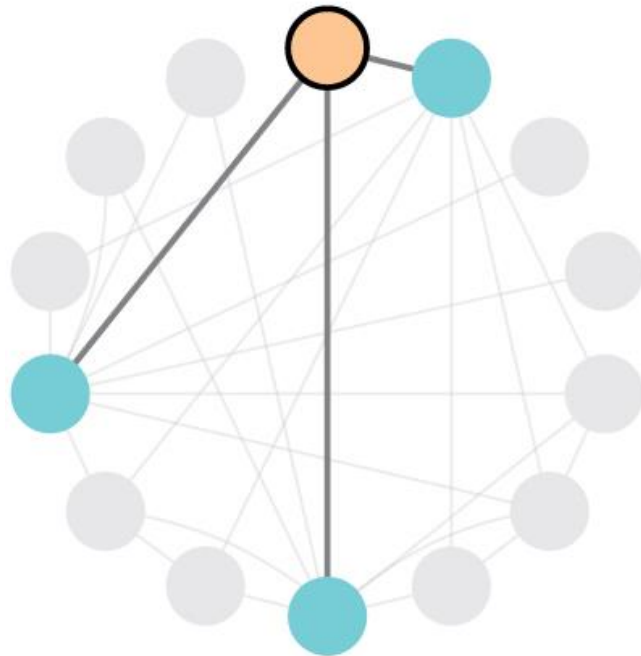
**Most nodes think blue is popular**



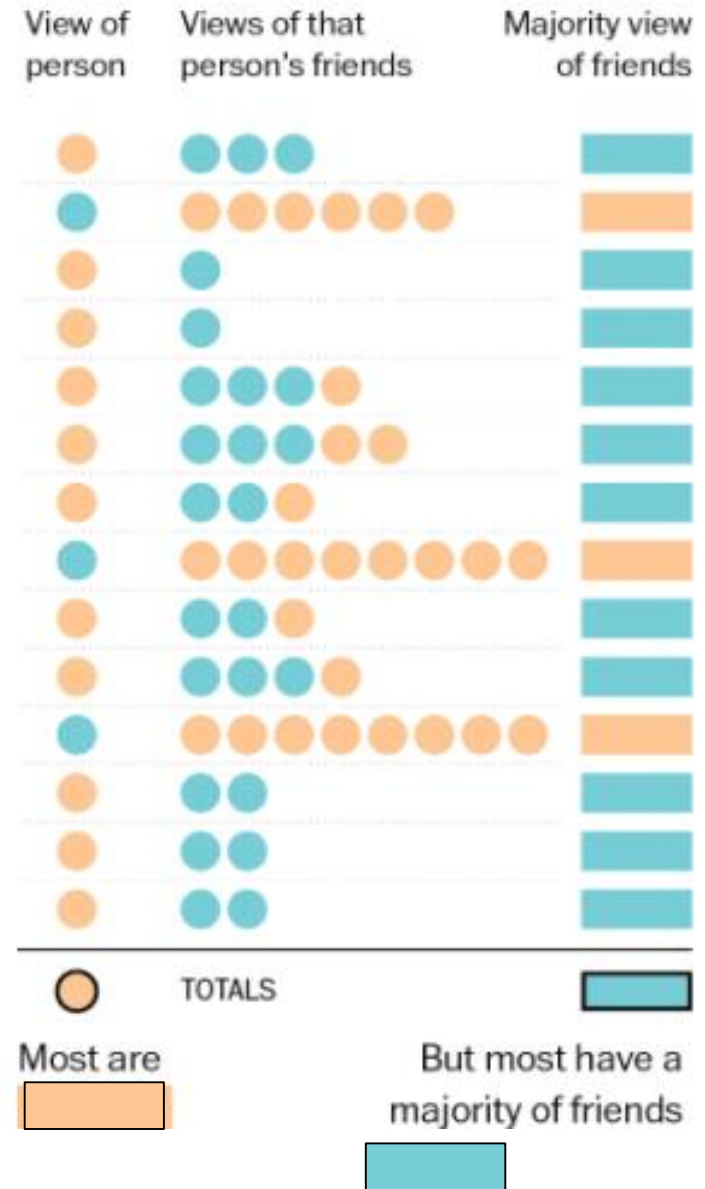
**Nodes think yellow is more popular  
Blue is not especially popular**

# Local vs global views

**100%** of this person's friends think baseball caps are trendy.



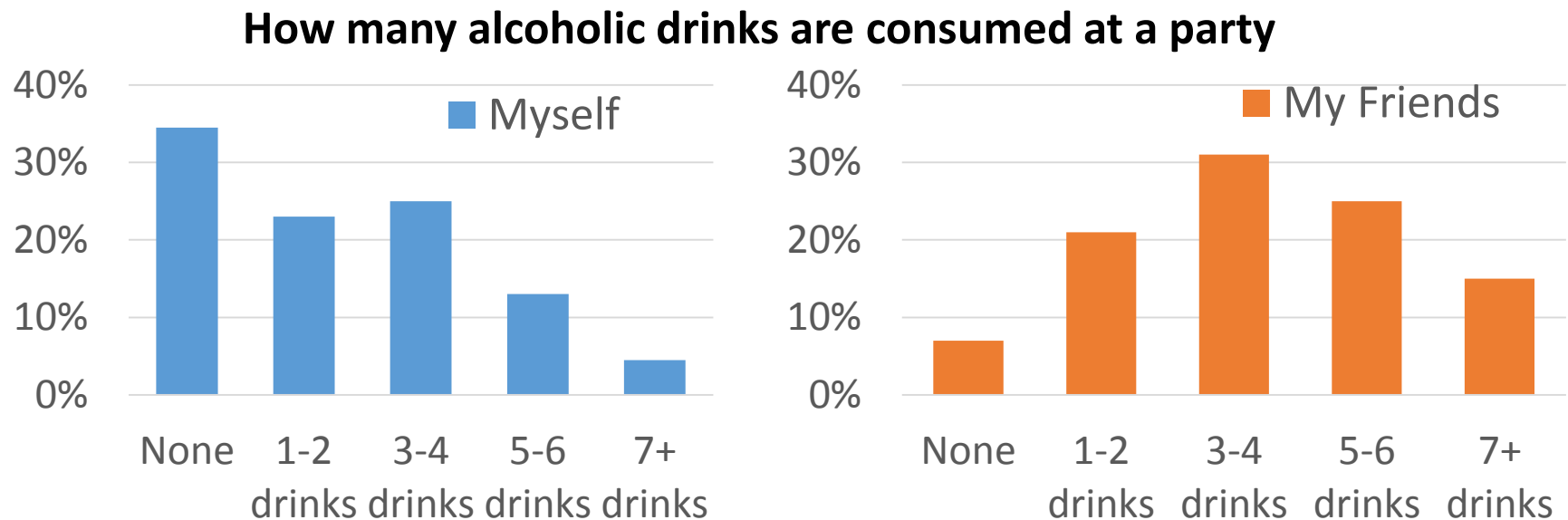
## What the network looks like to each person



K. Schaul, "A quick puzzle to tell whether you know what people are thinking", *Wonkblog, Washington Post*  
<https://www.washingtonpost.com/graphics/business/wonkblog/majority-illusion/>

# Networks distort local information

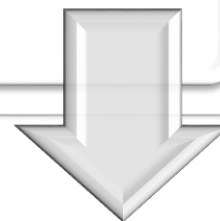
- Networks can systematically bias individual perceptions of what is common among peers
- Example: College students overestimate peers' alcohol use



# Outline

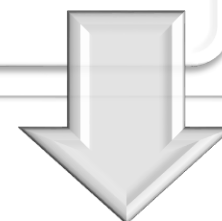
## Friendship paradox

- The many friendship paradoxes in networks
- Origins of paradoxes: a network science view



## Perception bias

- Friendship paradox in directed networks
- ... biases perceptions of popularity
- Twitter case study: Some hashtags appear more popular than they are



## Polling

- Estimate global popularity with perception bias
- ... with a limited budget



# Lots of paradoxes!

## Friendship paradox

You friends have more friends than you do, on average [Feld, 1991]

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## Generalized friendship paradox

You friends are more  $X$  than you are, on average [Hodas et al., 2013, Eom & Jo, 2014]

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## Strong friendship paradox

Most of your friends have more friends than you do [Kooti et al., 2014]

# Lots of paradoxes!

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## Generalized strong friendship

Most of your friends are more  $X$  than you are [Kooti et al, 2014]



## Majority illusion

Most of your friends have a trait, even when it is rare. [Lerman et al, 2016]

# How common is strong friendship paradox?

**Very common... almost everyone observes that most of their friends are more popular**

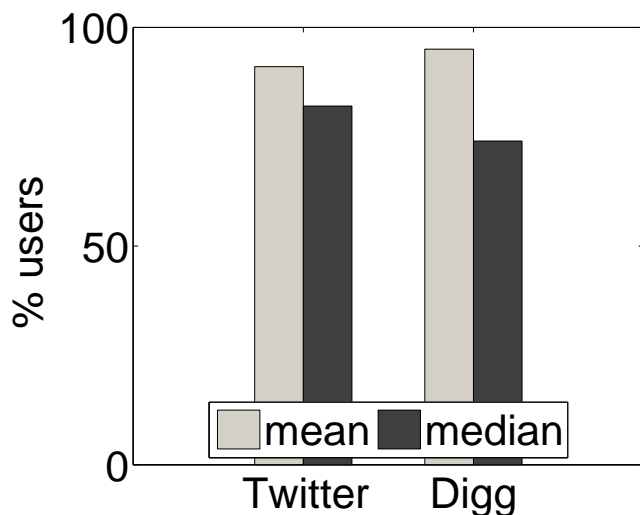
Network	Type	Nodes	Probability of paradox
LiveJournal	Social	3,997,962	84
Twitter	Social	780,000	98
Skitter	Internet	1,696,415	89
Google	Hyperlink	875,713	77
ProsperLoan	Social Finance	89,269	88
ArXiv	Citation	34,546	79
WordNet	Semantic	146,005	75



# But wait, there is more

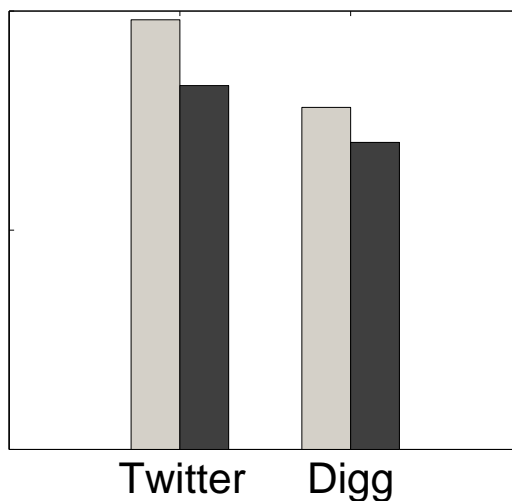
## Activity

You post less than most of your friends



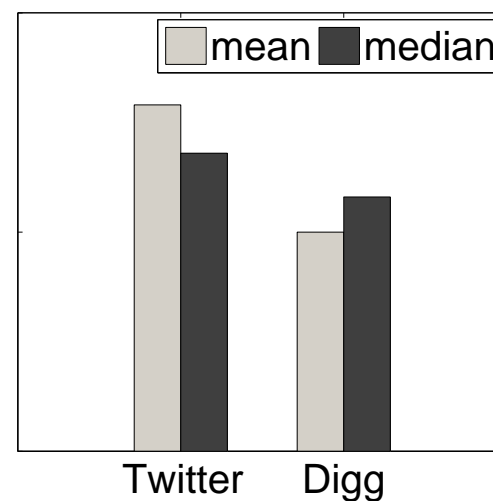
## Diversity

You see less diverse content than most of your friends



## Virality

You see less viral content than most of your friends



# Why?

## Friendship paradox

You friends have more friends than you do, on average [Feld, 1991]



## Generalized friendship paradox

You friends are more  $X$  than you are, on average [Hodas et al., 2013, Eom & Jo, 2014]

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# Different explanations

## Friendship paradox

You friends have more friends than you do, on average [Feld, 1991]



## Generalized friendship paradox

You friends are more X than you are, on average [Holme et al., 2017; Kim & Jo, 2014]

**STATISTICAL SAMPLING EXPLANATION**

## Strong friendship paradox

Most of your friends have more friends than you do [Kooti et al., 2014]



## Generalized strong friendship paradox

Most of your friends are more X than you are [Kooti et al., 2014]

**HIGHER ORDER NETWORK STRUCTURE**

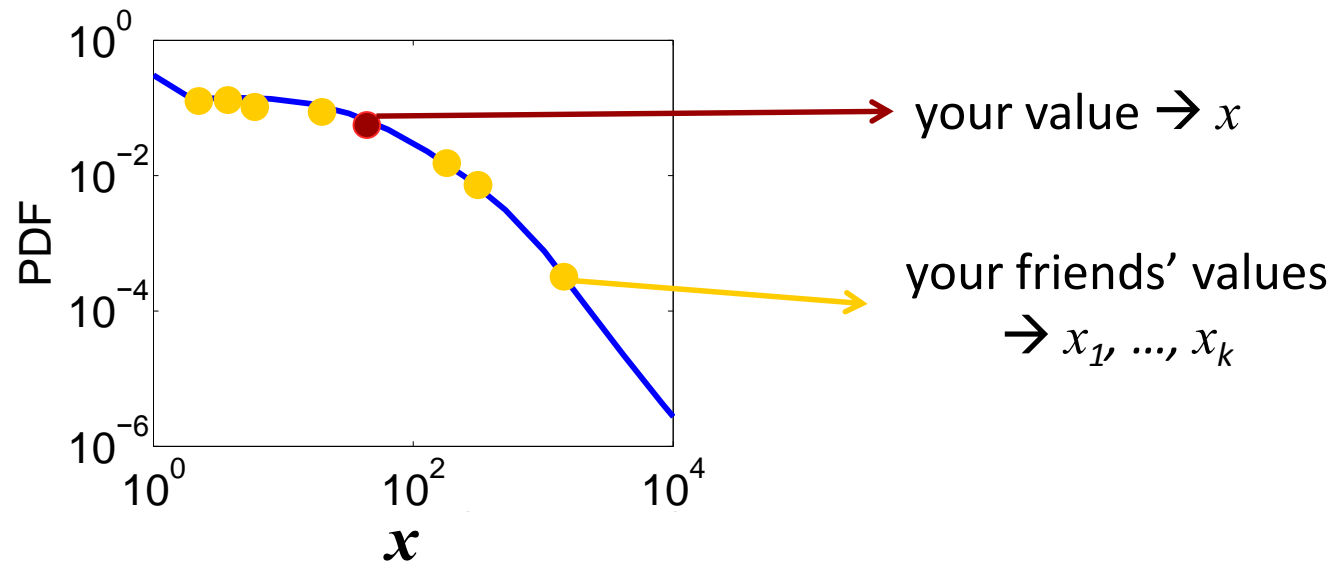


## Majority illusion

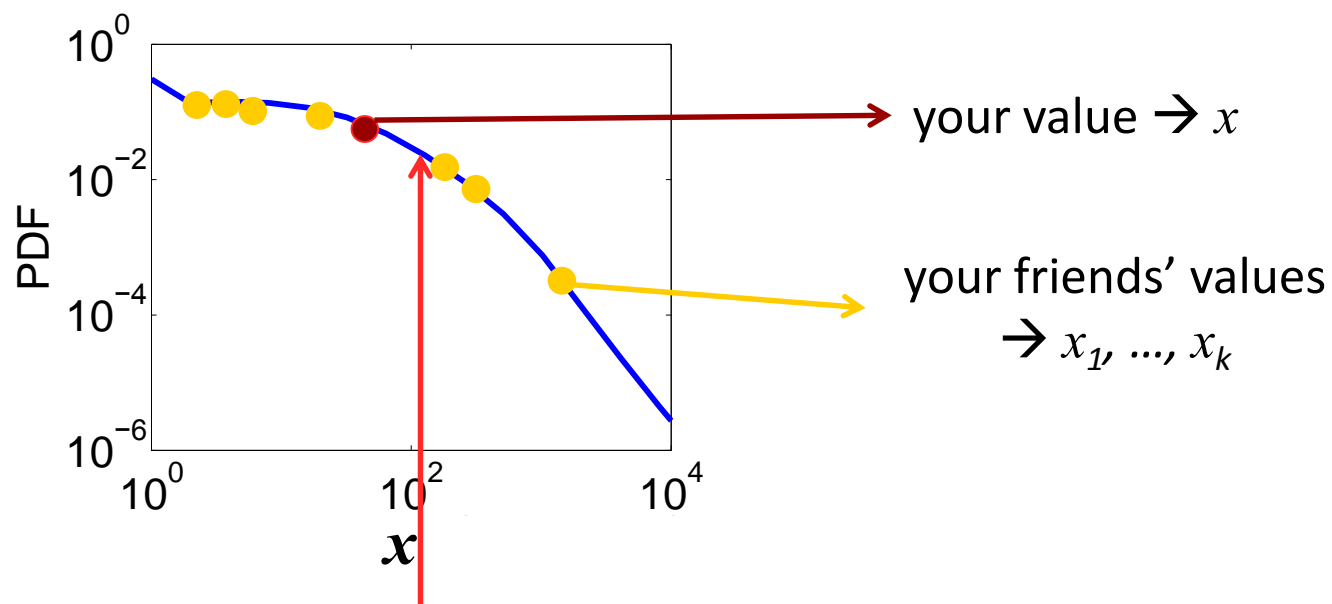
Most of your friends have a trait, even when it is rare. [Lerman et al, 2016]



# Friendship paradox as a byproduct of sampling from a heterogeneous distribution



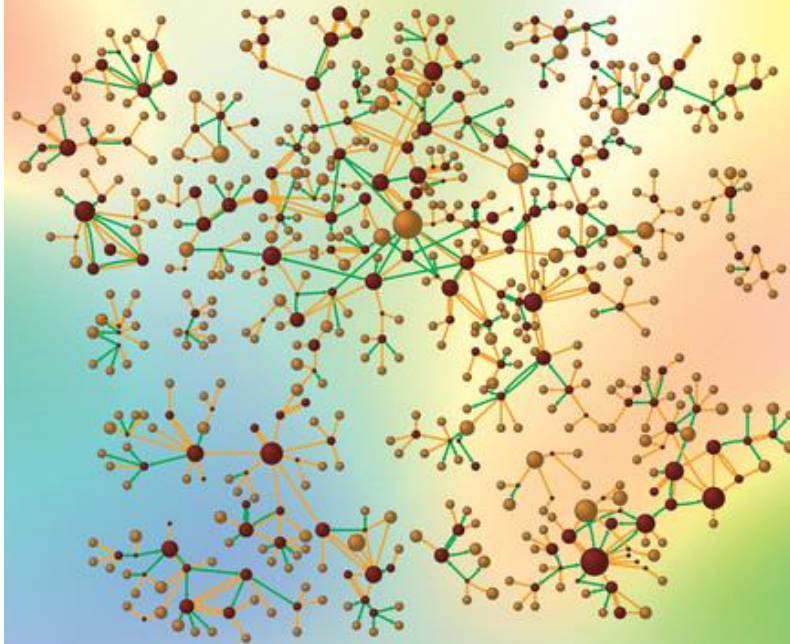
# Friendship paradox as a byproduct of sampling from a heterogeneous distribution



$$\text{mean} = \frac{1}{k} (x_1 + \dots + x_k) > x \quad \text{Sample mean grows with } k$$



# NETWORK SCIENCE



NATIONAL RESEARCH COUNCIL  
OF THE NATIONAL ACADEMIES

# Strong friendship paradox is a network effect

- To explain strong friendship paradox, need to account for network structure
- Building blocks of network structure
- $d$ K series framework represents network structure as the joint degree distribution of subgraphs of up to  $d$  nodes

P. Mahadevan, D. Krioukov et. al., *ACM SIGCOMM Comp. Comm. Rev.* 36 135–146 (2006)

## First-order structure (1K)

- Node degree distribution

## Second-order structure (2K)

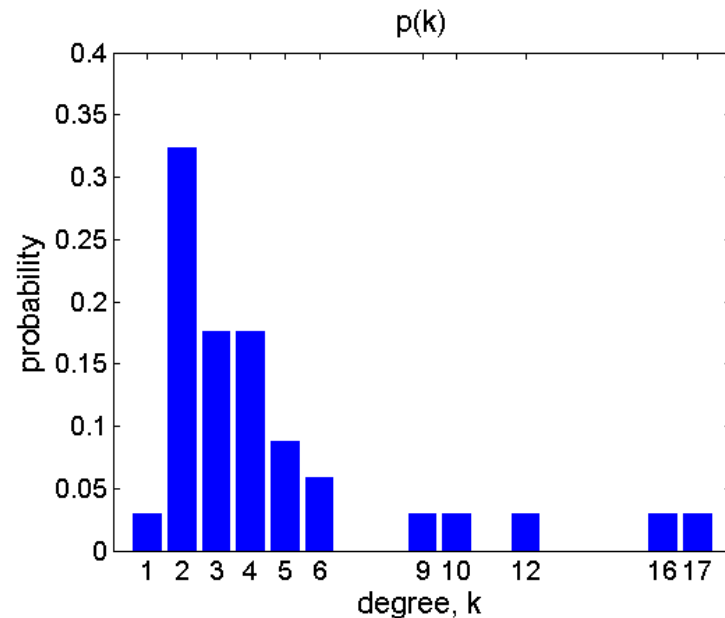
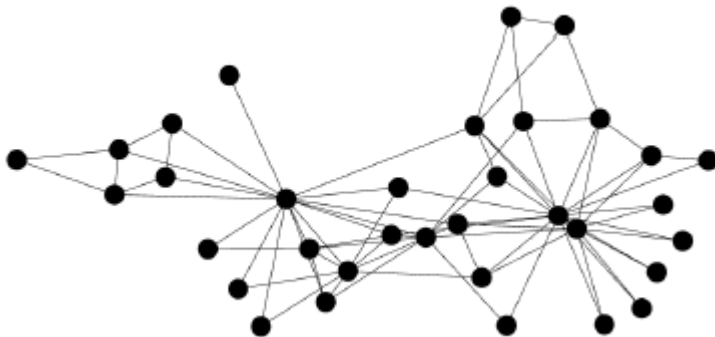
- Pair degree distribution

## Third-order structure (3K)

- Triplet degree distribution

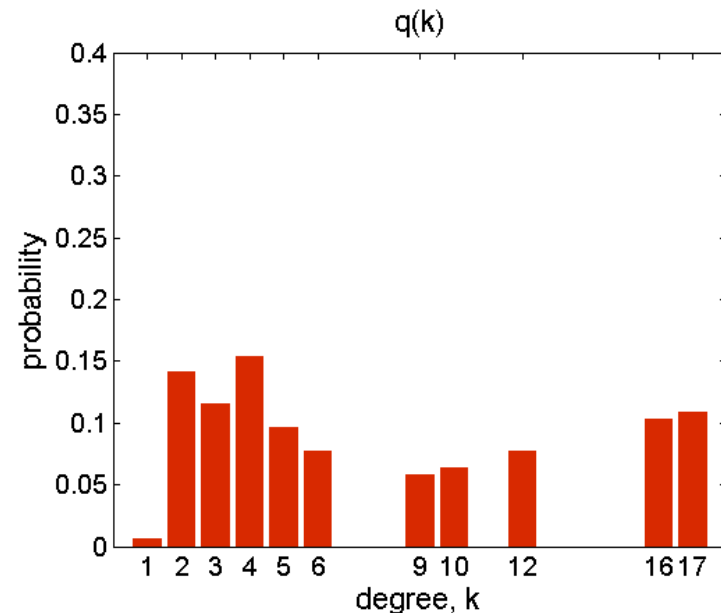
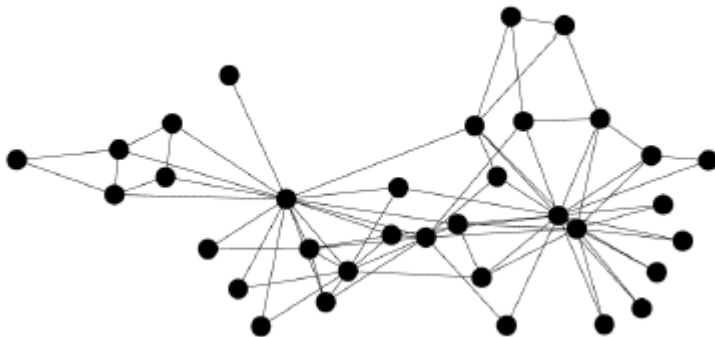
## First-order (1K) structure

- Node degree distribution  $p(k)$ 
  - Probability that a randomly selected *node* has degree  $k$ .
- Any heterogeneous degree distribution (variance  $> 0$ ) will lead to a (weak) friendship paradox

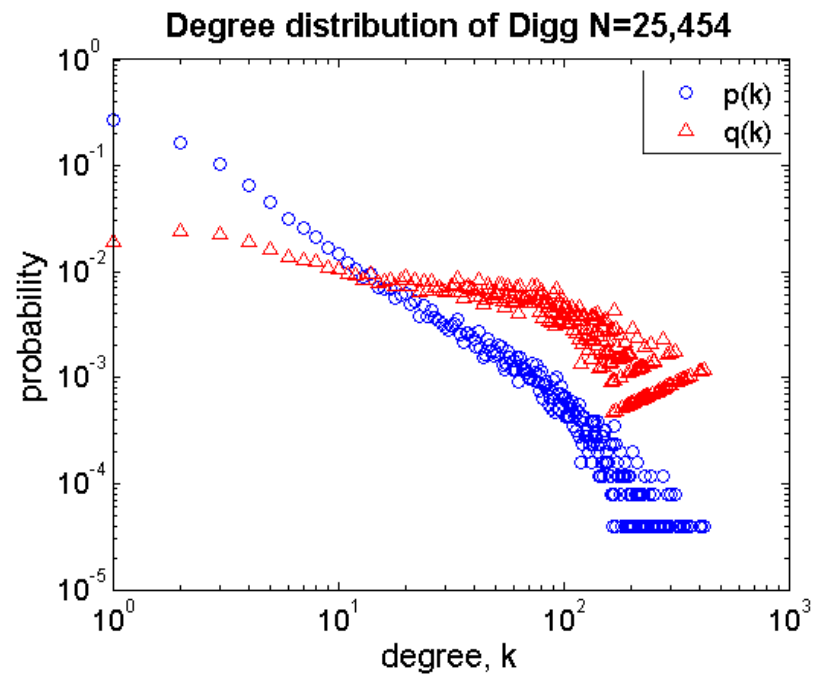
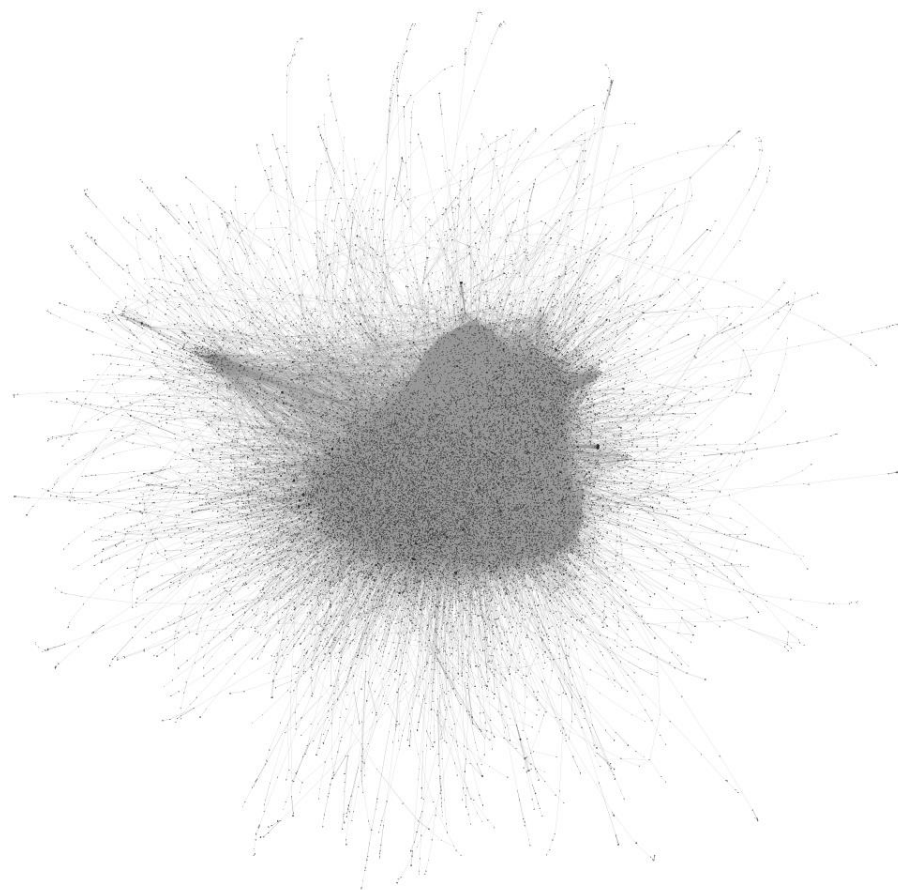


## First-order (1K) structure

- Neighbor degree distribution  $q(k) \sim kp(k)$
- Probability that a randomly selected *neighbor* has degree  $k$ .
- Any heterogeneous degree distribution (variance  $> 0$ ) will lead to a (weak) friendship paradox



# Digg social network



## Second-order (2K) structure

- ... Nodes do not link at random
- Joint degree distribution of connected pairs of nodes  $e(k, k')$ 
  - Probability that a randomly selected *edge* links nodes with degrees  $k$  and  $k'$ .
- **Degree assortativity**  $r_{2k}$ 
  - MEJ Newman, Assortative Mixing in Networks, *Phys Rev Lett* 89 208701 (2002)

**disassortative**



**assortative**



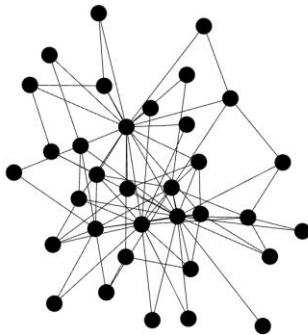


## Third-order (3K) structure

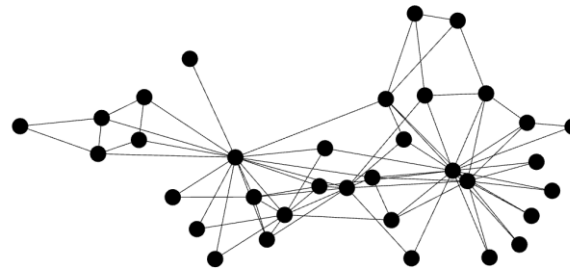


- ... nodes do not link to random neighbors
- **Neighbor assortativity:** neighbors tend to have similar (or dissimilar) degrees,  $r_{3k}$
- Networks can have the same 1K and 2K structure but different 3K structure
- Wu, Percus & Lerman, Neighbor Degree Assortativity in Networks, *in preparation*

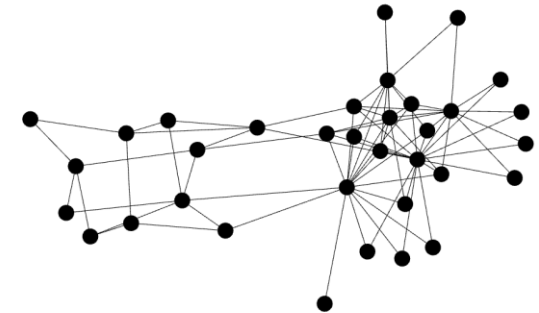
negative  
neighbor assortativity



original network

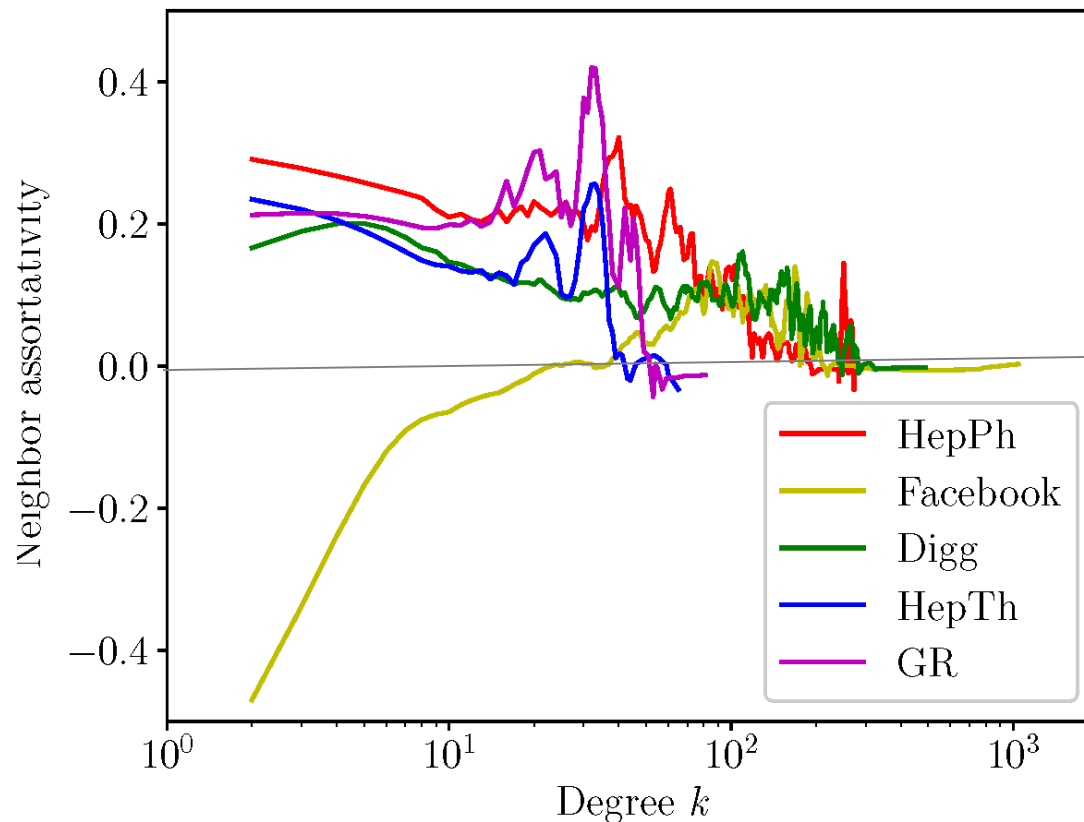


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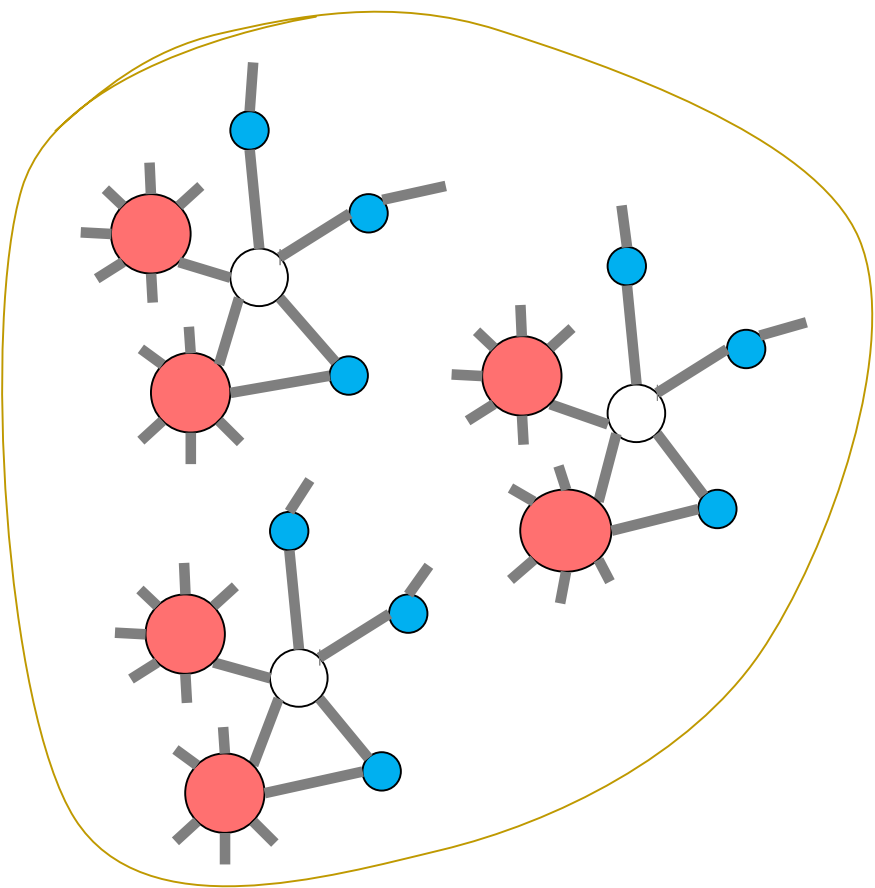
# Real-world networks have third-order structure

Degree correlations among node's neighbors in real-world networks are often large

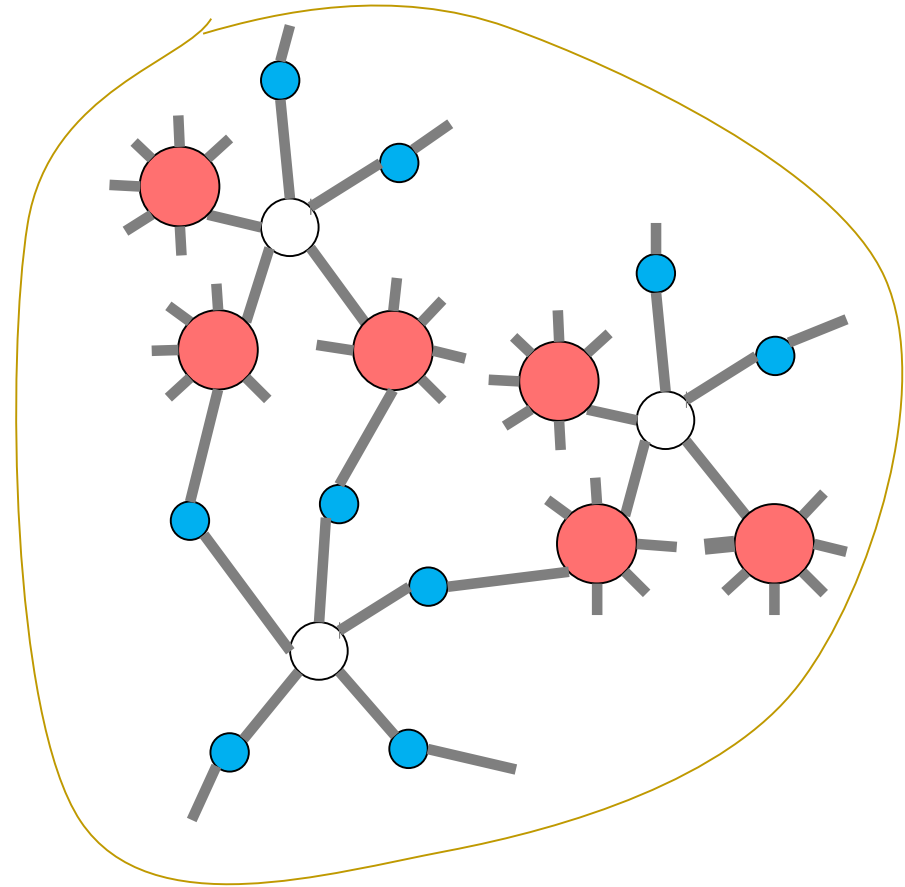


# Third-order structure enhances paradoxes

Neighbors' degrees are not correlated\*



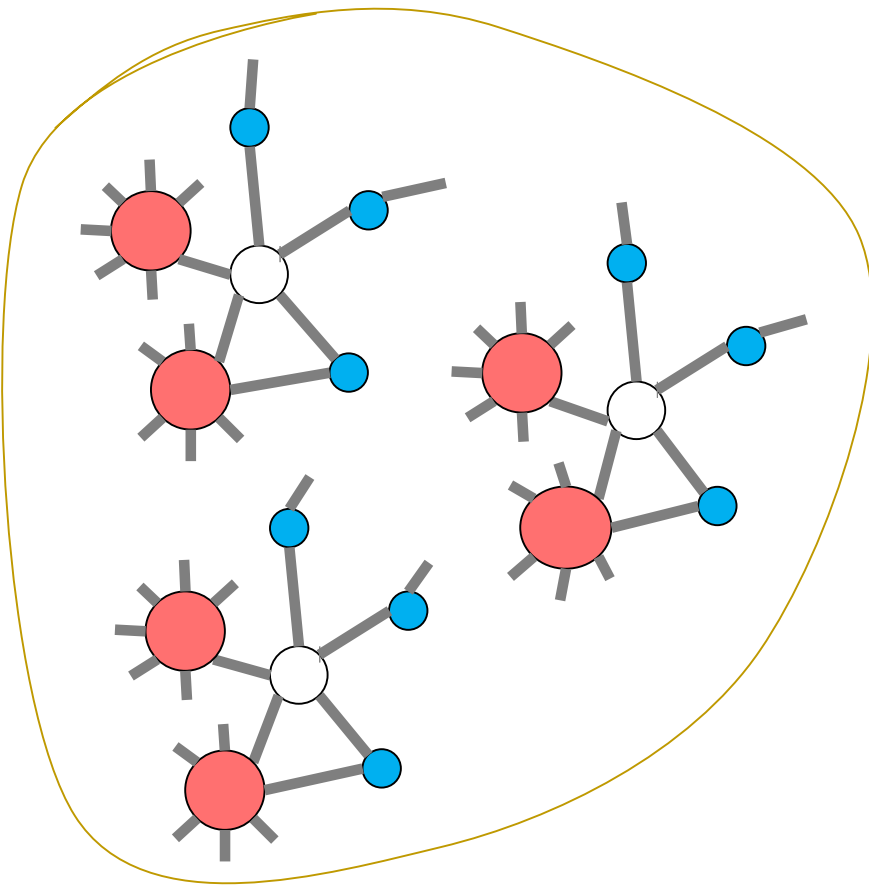
Neighbors' degrees are correlated\*



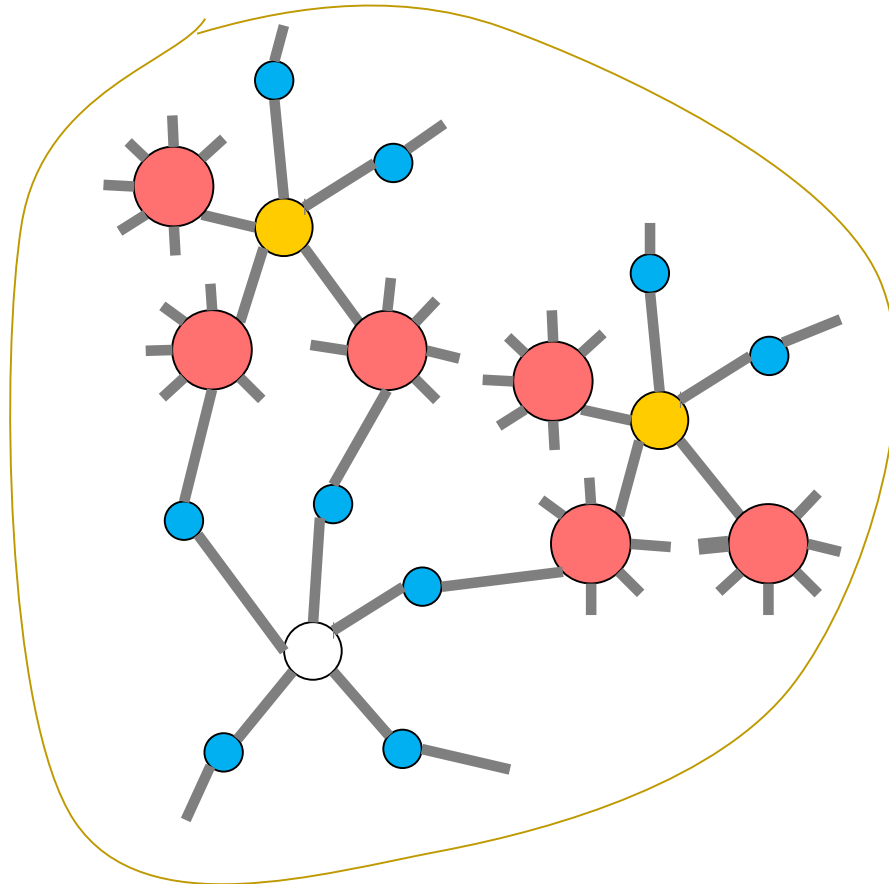
\*same 1K and 2K structure

# Third-order structure enhances paradoxes

Neighbors' degrees are not correlated\*



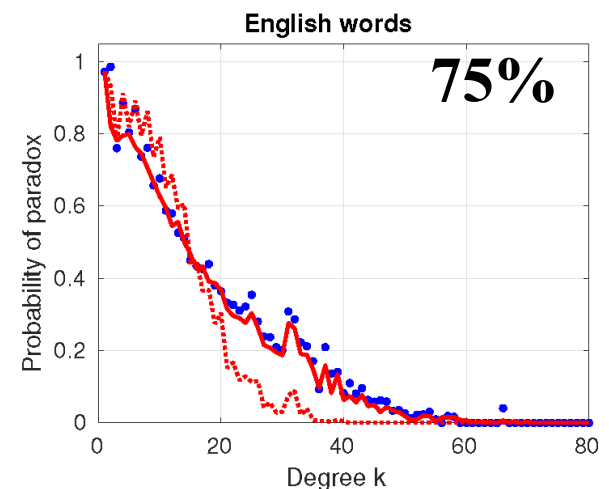
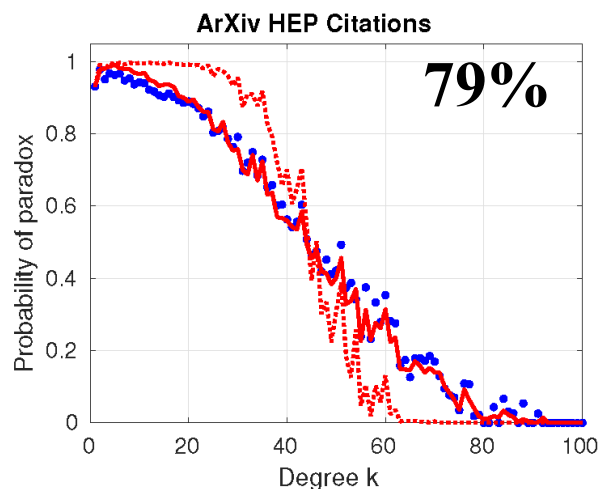
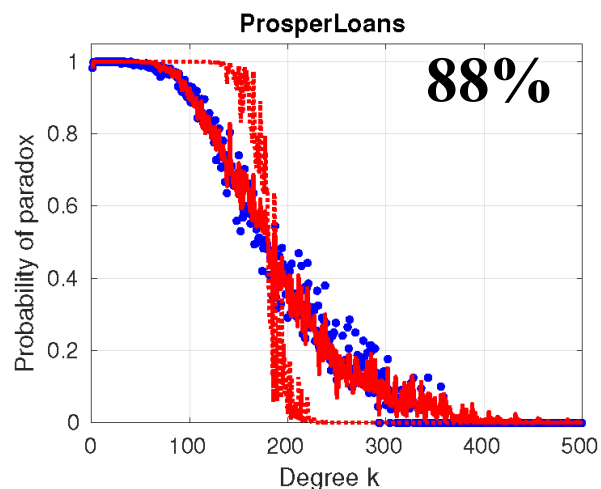
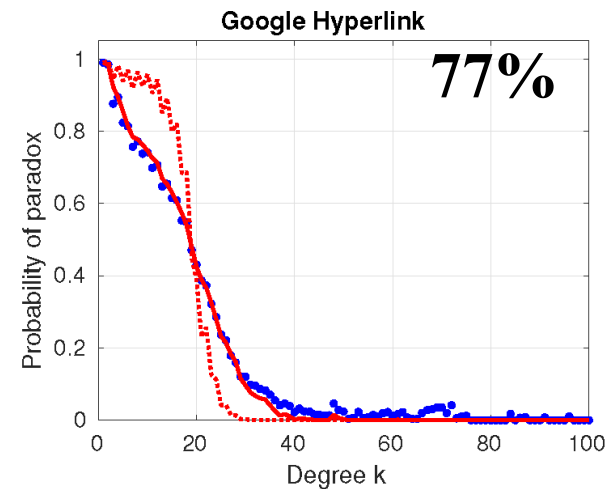
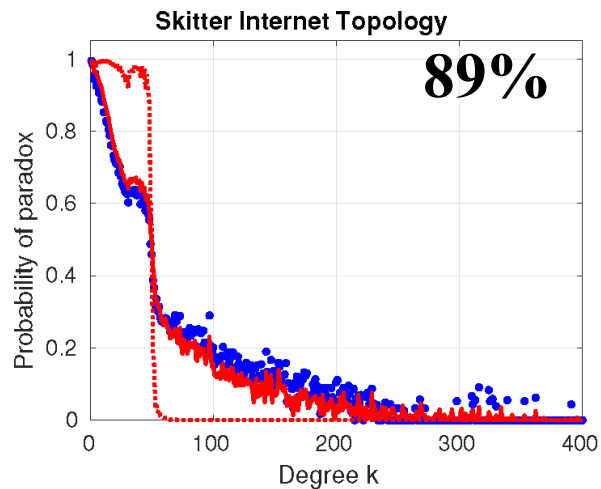
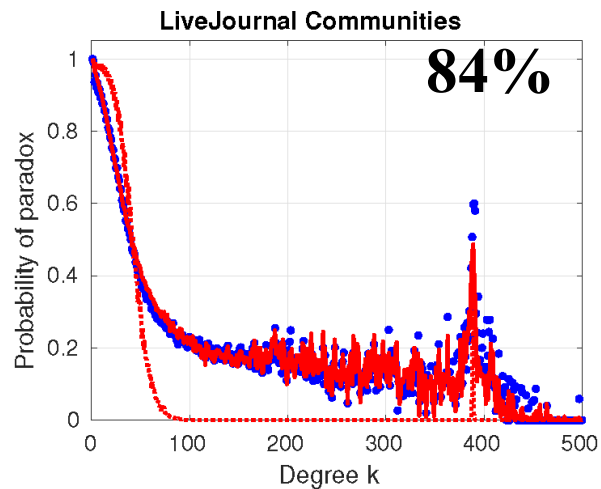
Neighbors' degrees are correlated\*



\*same 1K and 2K structure



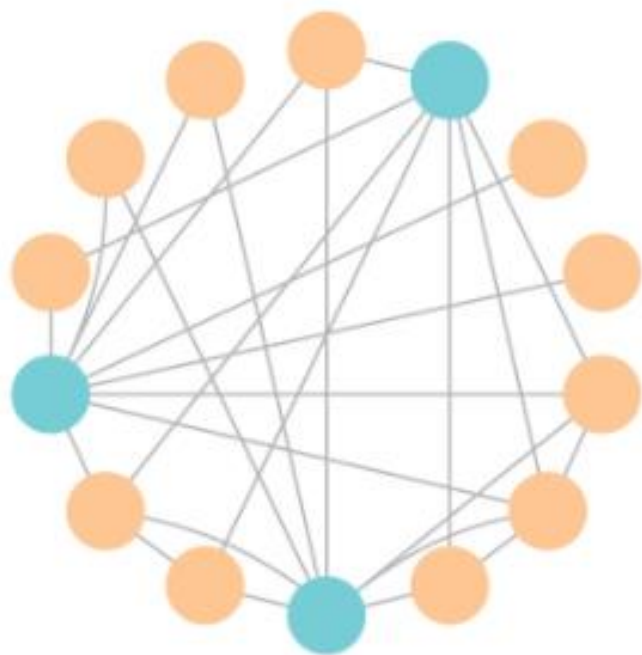
# Fraction of degree- $k$ nodes experiencing the paradox in real-world networks • ; predictions of the 2K model - - - and the 3K model —.



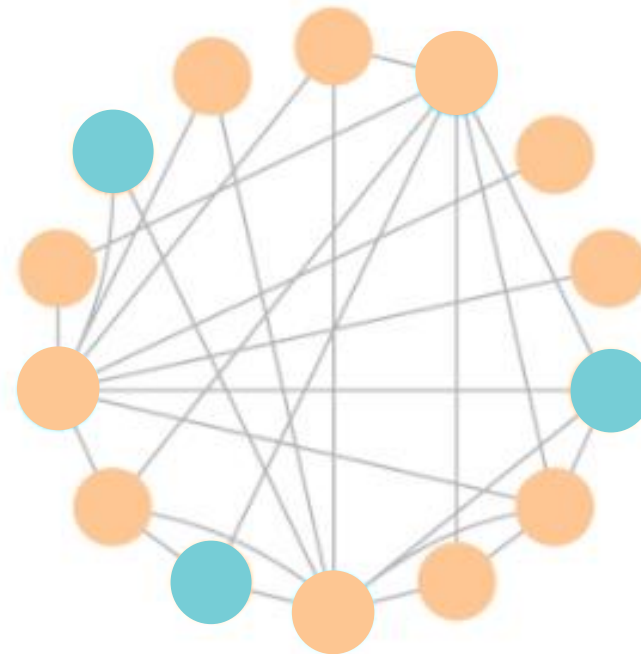
[Wu et al (2017) "Neighbor-neighbor correlations explain measurement bias in networks" *Scientific Reports* 7]

# From friendship paradox to “majority Illusion”

Nodes have a binary trait: active/not, yellow/blue, heavy drinker/teetotaler, ...



**Many think that blue is common**

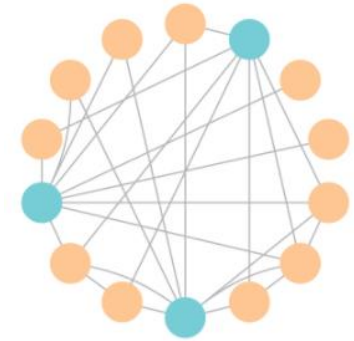


**Blue does not appear common**

# Network structure amplifies majority illusion

More nodes will think that blue is very common when:

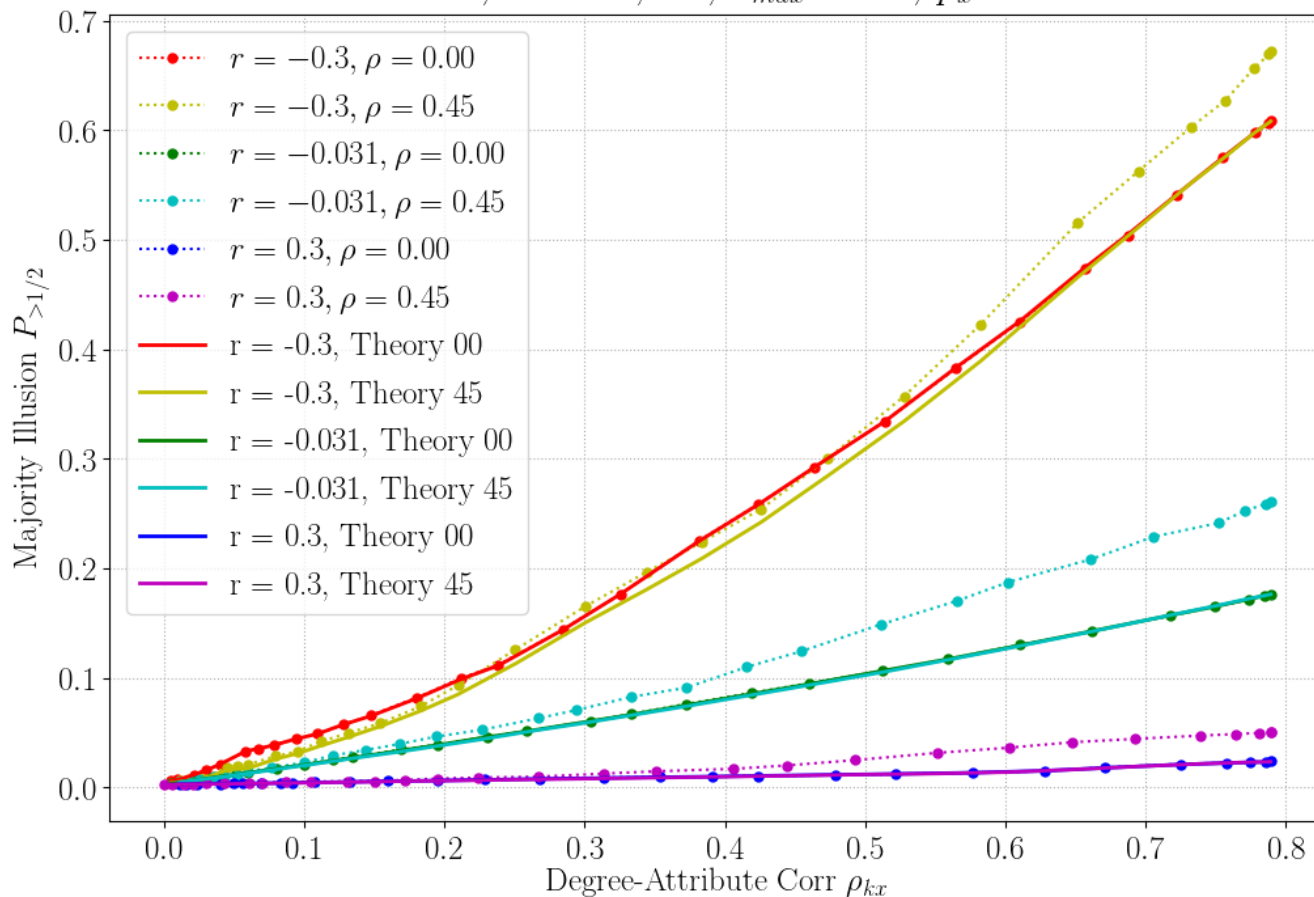
- Higher degree nodes are more likely to be blue: degree-trait ( $k$ - $x$ ) correlation
- High degree nodes link to low degree nodes: degree disassortativity ( $r_{2k} < 0$ )
- Neighbors tend to have similar degree: neighbor assortativity ( $r_{3k} > 0$ )



# Network structure amplifies majority illusion

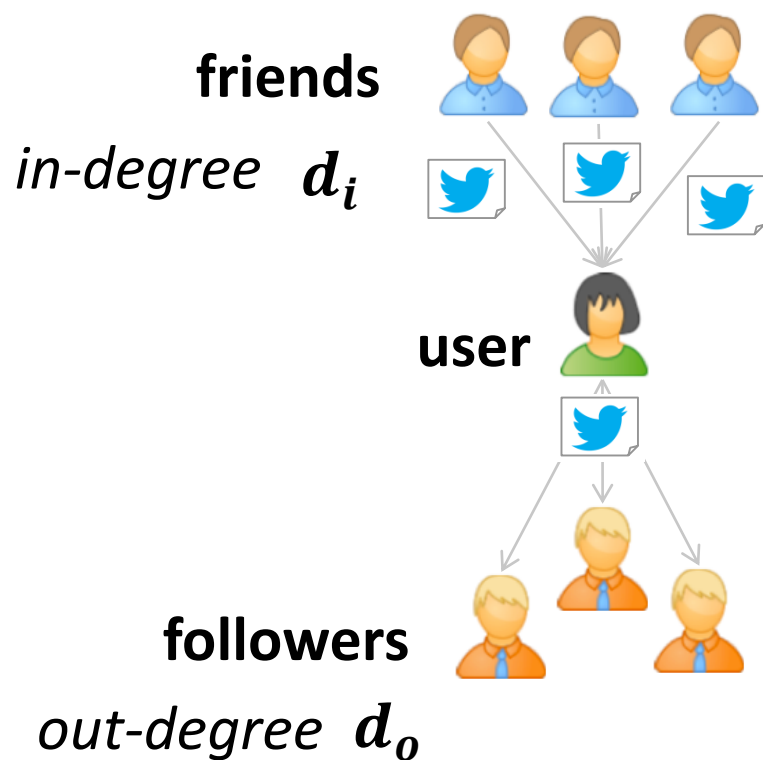
Fraction of nodes experiencing the majority illusion in a synthetic network with 0.5% active nodes;

$$\alpha = -2.1, n = 100,000, k_{max} = 876, p_x = 0.005$$

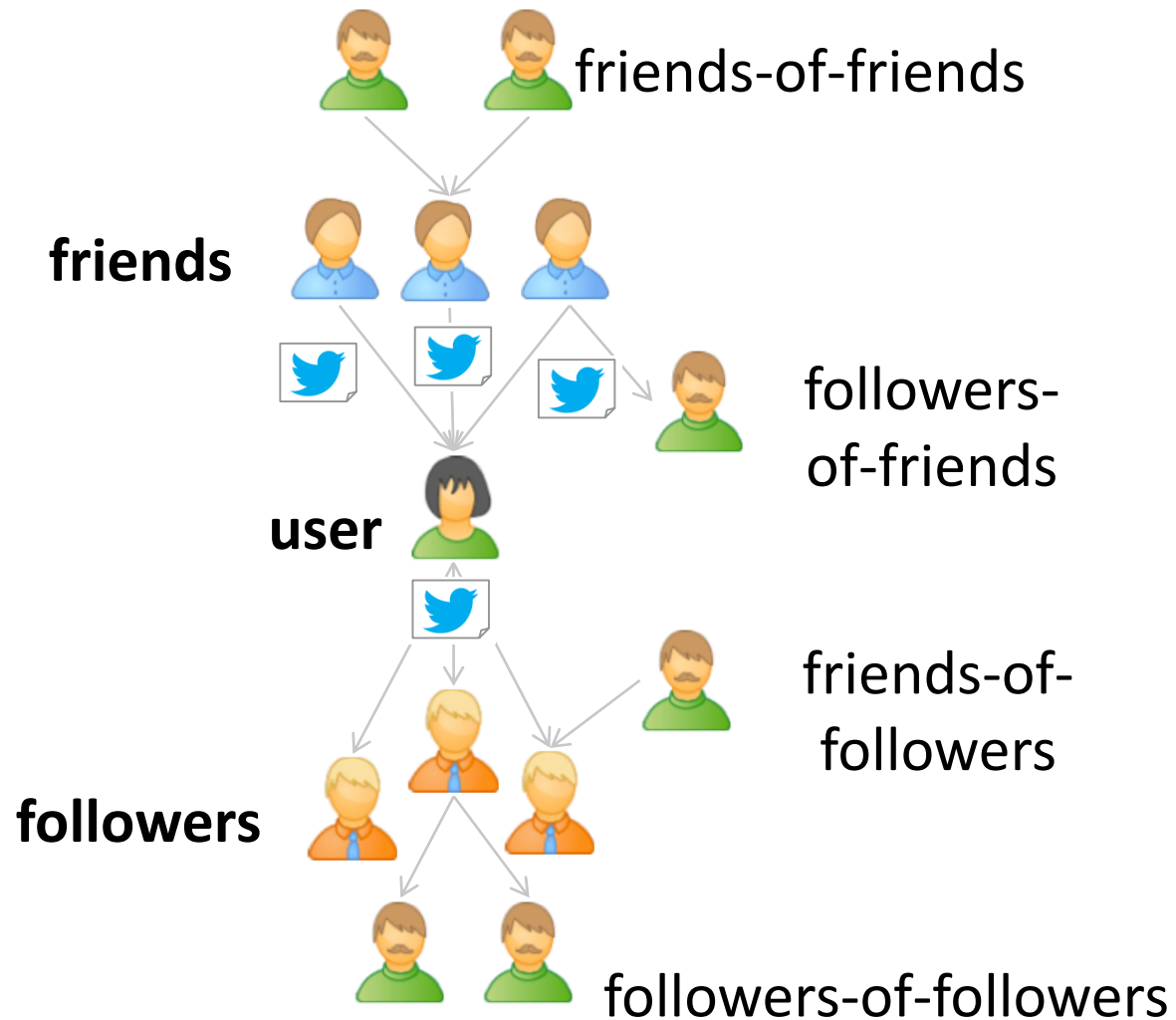




# Friendship paradox in directed networks

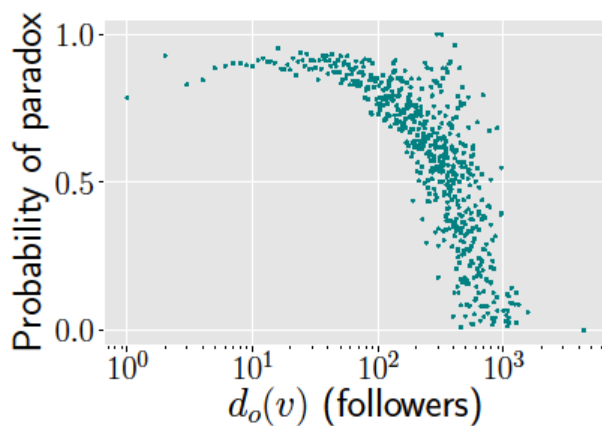


# Friendship paradox in directed networks

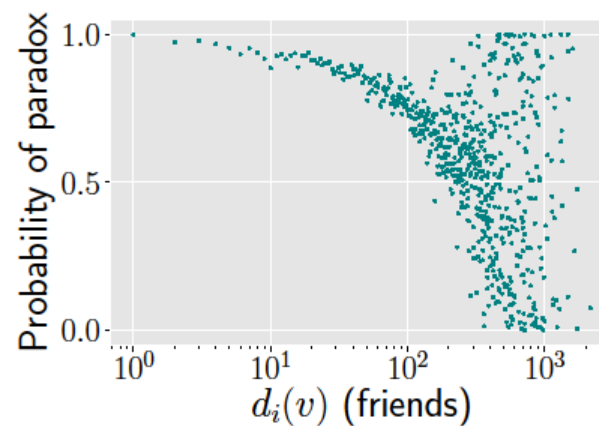


# Probability a node experiences a paradox

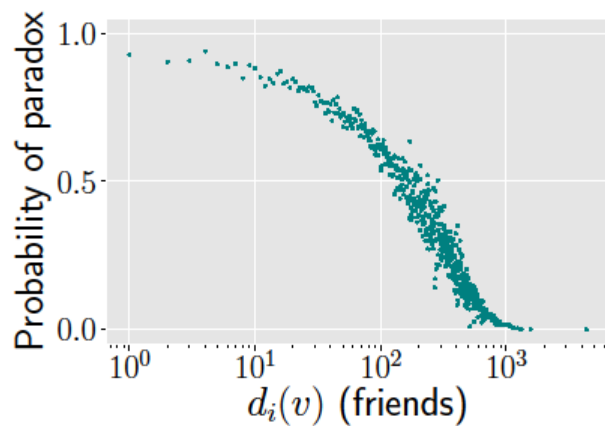
friends have more followers



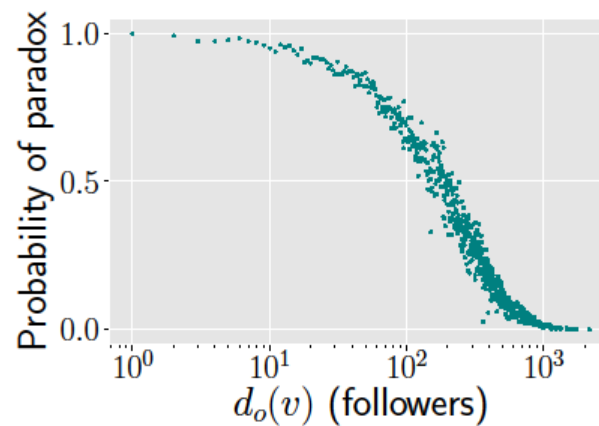
followers have more friends



friends have more friends



followers have more followers



## Local perception bias

- Popularity of a (binary) attribute: probability a random node  $v$  has value  $f(v)$

$$\mathbf{E}\{f(v)\}$$

- Local perception of node  $v$  about popularity of an attribute  $f$  is the fraction of her friends with attribute

$$q(v) = \frac{\sum_{u \in \text{friends}(v)} f(u)}{d_i(v)}$$

- Local perception bias: nodes perceive the attribute  $f$  to more popular than it actually is

$$\mathbf{E}\{q(v)\} \geq \mathbf{E}\{f(v)\}$$

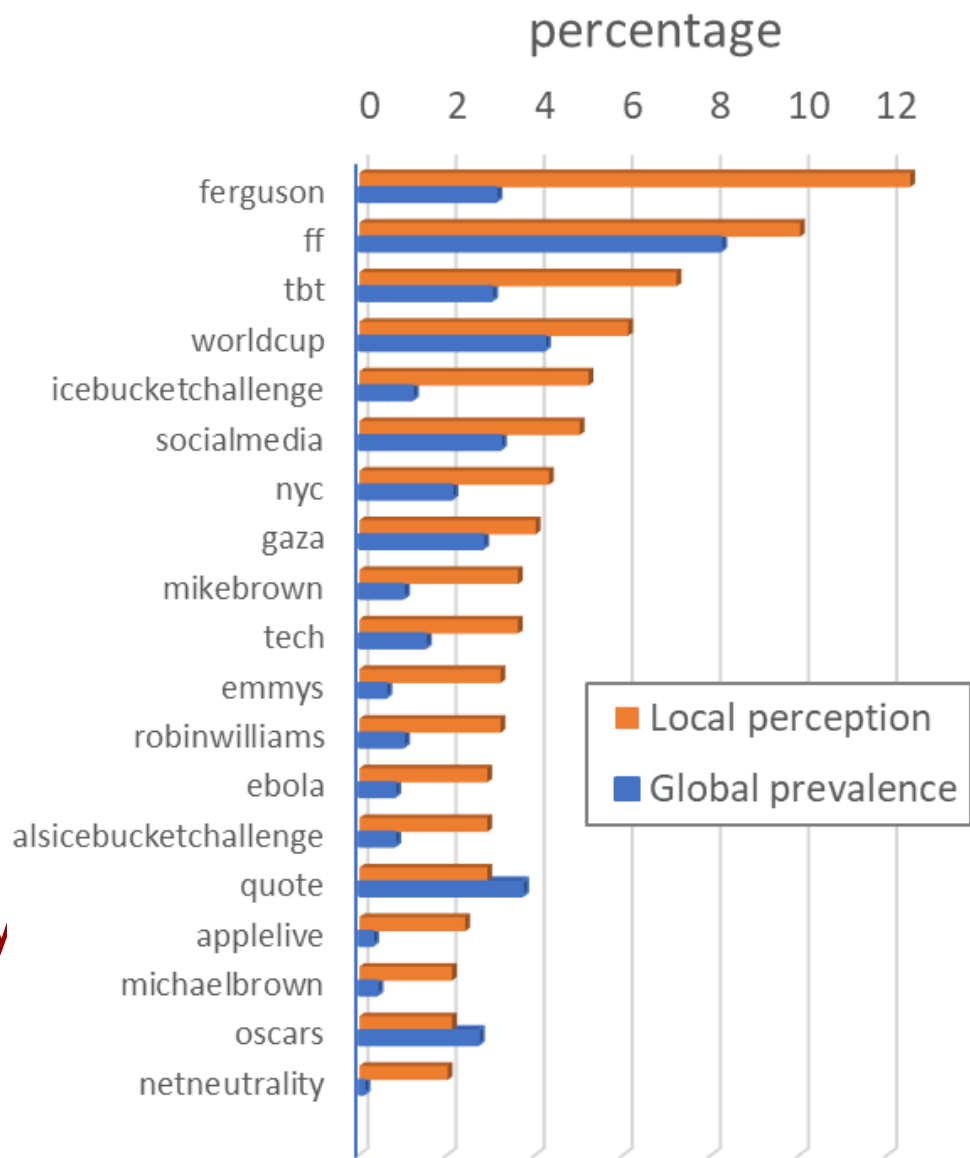


# Global popularity vs local perception on Twitter

## Twitter data

- Time period
  - Summer 2014
- Network
  - 5K users + tweets
  - Their 600K friends + tweets
- Hashtags
  - 18M hashtags
  - Focus on 1K most popular hashtags, used by >1K people

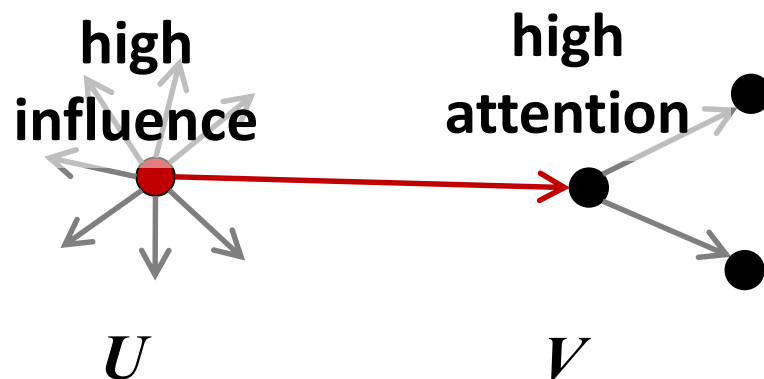
Compare perceived popularity of hashtags to their actual popularity



# Conditions for local perception bias

Local bias exists if:

- Higher out-degree (high influence) tend to have the attribute.
- Lower in-degree nodes (high attention) tend to follow nodes with attribute.



**Theorem:**  $(\mathbf{E}\{q(v)\} \geq \mathbf{E}\{f(v)\})$  if

$$\text{Cov}\{f(X), d_o(X)\} \geq 0 \quad \text{and,}$$

$$\text{Cov}\{f(U), \mathcal{A}(V) | (U, V) \sim \text{Uniform}(E)\} \geq 0.$$

# Polling



# What is the right question to ask in a poll?

Estimate the true prevalence of an attribute through polls

- Estimate the fraction of liberals vs conservatives, heavy drinkers vs teetotalers, people who used a hashtag vs not, ...
- ...with limited budget  $b$

## Polling:

1. Intent Polling (IP): [ $b$  random nodes] Will you vote for X?
2. Node Perception Polling (NPP): [ $b$  random nodes] What fraction of your friends will vote for X?
3. **Follower Perception Polling (FPP): [ $b$  random followers]**  
**What fraction of your friends will vote for X?**
  - aggregates perceptions of more people



## Bias & variance of FPP

- Bias of the polling estimate (error)  $T$

$$\text{Bias}(T) = E\{T\} - E\{f(X)\} = \frac{\text{Cov}(f(X), d_o(X))}{\hat{d}}$$

- Variance is bounded by  $\lambda_2$ , second largest eigenvalue of the symmetrized adjacency matrix of the network

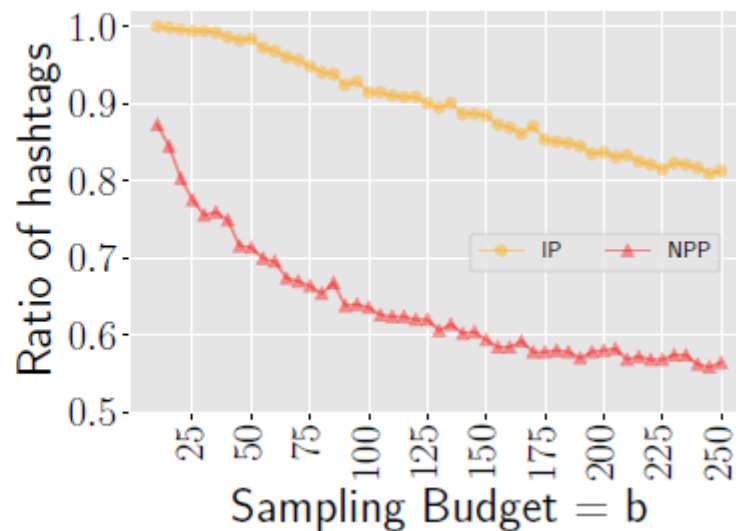
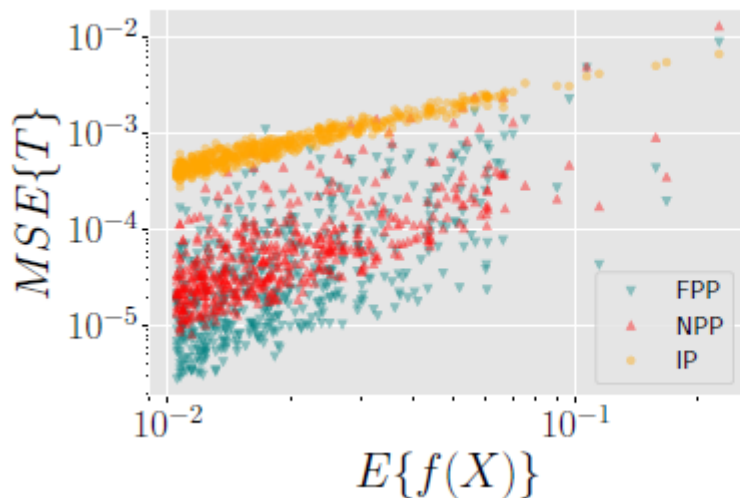
- Mean squared error of the polling estimate  $T$ :

$$\text{MSE}\{T\} = E\left\{[T - E(f(X))]^2\right\} = \text{Bias}\{T\}^2 + \text{Var}\{T\}$$

# FPP polling algorithm is more efficient

When used to estimate the true popularity of Twitter hashtags, FPP has lower variance and MSE.

For a given budget, i.e., number of nodes sampled, it outperforms other polling methods on many hashtags.



## To summarize

- Network structure can systematically bias local perceptions
  - Making a rare attribute appear far more common than it is, under some conditions
- Open questions: What is the impact of network bias on
  - Collective dynamics in networks, e.g., contagious outbreaks
  - Network control and intervention
  - Psychological well-being
    - Your friends are happier than you are (Bollen et al. 2016)
    - Your co-authors are more prestigious than you are (Eom & Jo 2014)
    - Social comparison theory



# THANK YOU!

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Questions?

[lerman@isi.edu](mailto:lerman@isi.edu)