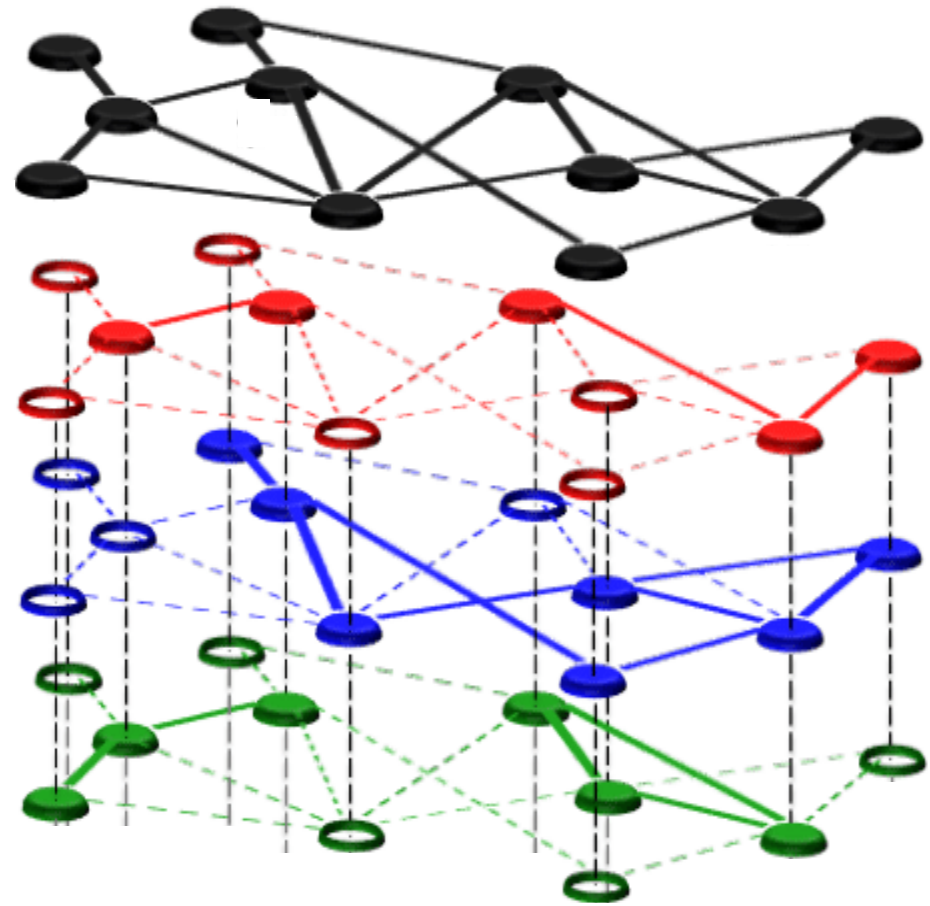
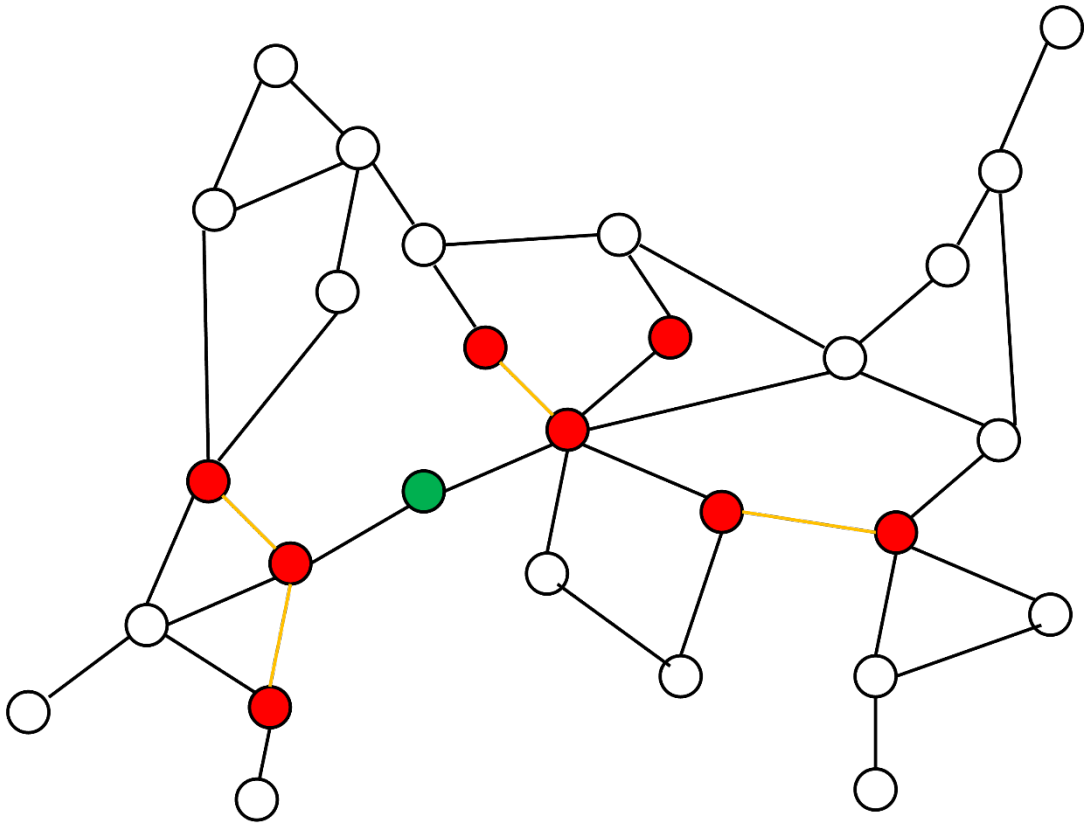


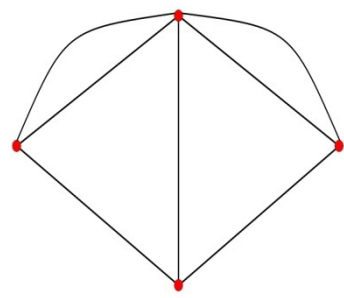
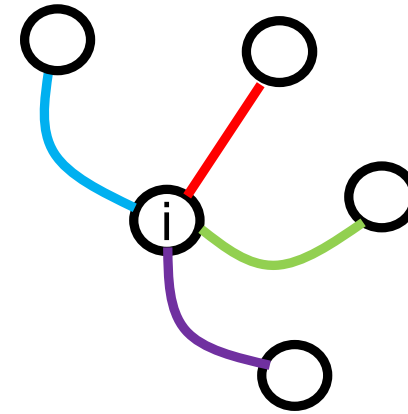
# Multiplexing and Diffusion



Chandrasekhar, Chaudhary, Golub, Jackson

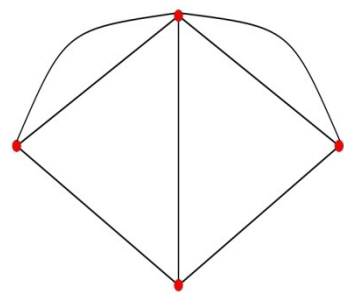
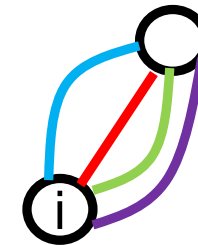
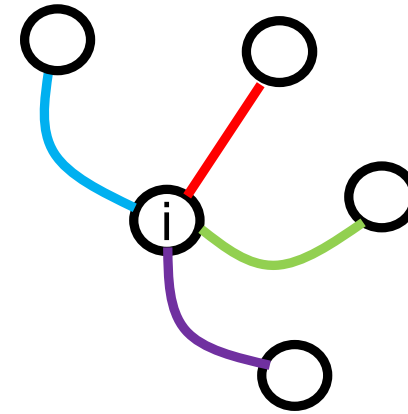
# “Multiplexed” Networks

- Compare two people:
  - One has different friends for different purposes
    - friends from work
    - friends they go to dinner with
    - friends they borrow money from
    - friends they give advice to...

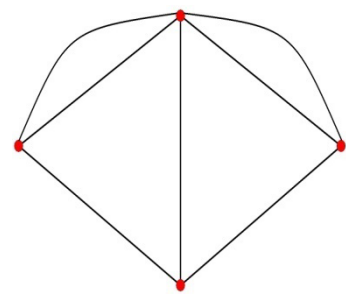


# “Multiplexed” Networks

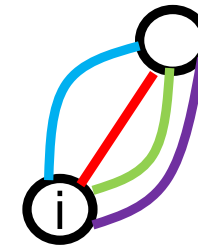
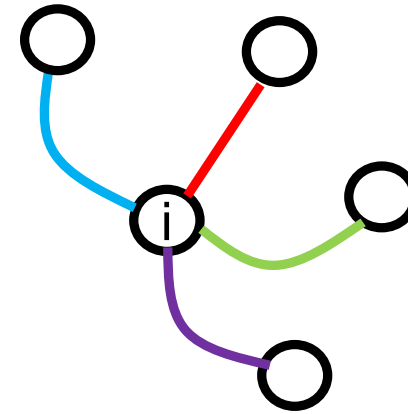
- Compare two people:
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# “Multiplexed” Networks

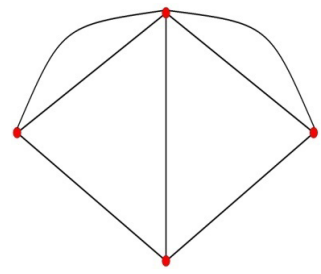


- Compare two people:
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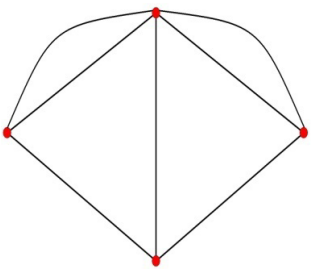
- Which one is more likely to learn/diffuse information?
- Which one is more likely to adopt a new technology, adopt a new behavior?

# Multiple Layers of Relationships



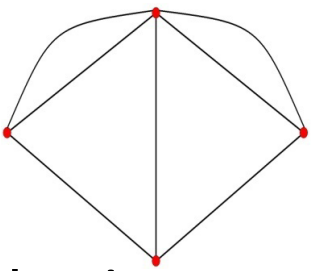
- Different types of relationships are correlated, but distinct
- Interacting with more people can help diversify access to information, new technologies
- But, more likely to interact with someone you do many things with, and can get more pressure from them...

# Questions



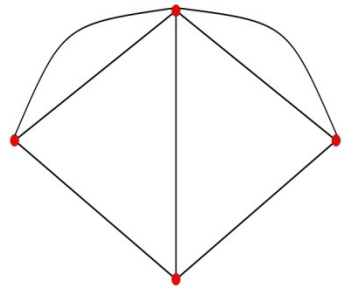
- Are there systematic patterns in how different layers of relationships overlap - “multiplexing”?
- How does multiplexing affect diffusion?
- How does the effect of multiplexing depend on what is being diffused?
  - simple contagion: spread of idea or disease
  - complex contagion: technology adoption, norms of behavior

# Multiplex Literature



- **Multiplexing in data**, surveys: Simmel 1908, Wasserman Faust 1994, Boccaletti et al. 2014, Kivella et al. 2014, Dickison et al. 2016
- Formation: Billand, Bravard, Joshi, Mahmud, Sarangi 2023, San Roman 2024
- Cooperation: Atkisson et al. 2020, Cheng, Huang, Xing 2021
- Games/Peers: Chen, Zenou, Zhou 2018, Walsh 2019, Zenou, Zhou 2023, Jackson, Zenou, Zhou 2024
- **Diffusion**: Yagan, Gligor 2012, Bianconi, Radicchi 2016, Kobayashi, Onaga 2023, Larson Rodriguez 2023

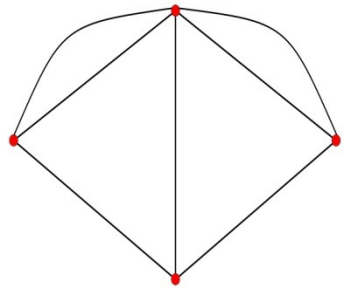
# Background Motivation



- BBCDJK 2024: Introduction of microfinance changes social networks
- Not only borrowing/lending networks, but also advice networks disappear

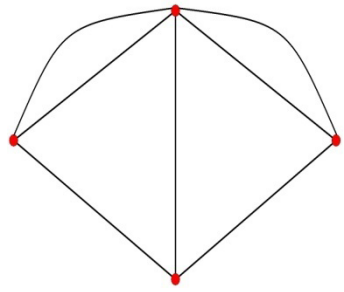


# Background - Microfinance



- Karnataka India 75 villages:
  - 43 villages people were offered microfinance loans
  - 32 controls (no loans offered)

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Banerjee, Chandrasekhar, Duflo, Jackson (Science 2013, Restud 2019)

Banerjee, Breza, Chandrasekhar, Duflo, Jackson, Kinnan (Restud, 2024)

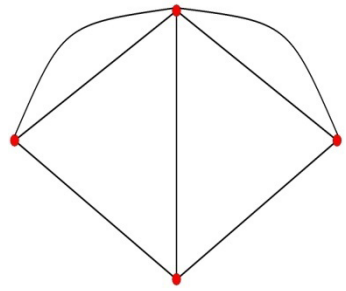


Karnataka



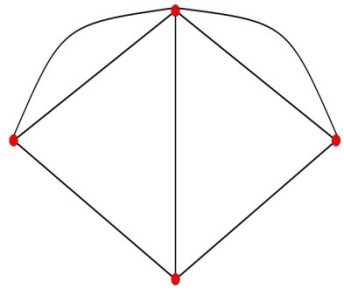


# Timeline



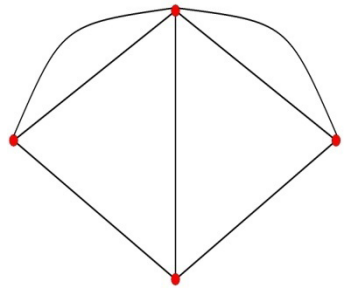
- 2006 We surveyed 75 villages that the bank intended to enter
- 2007-2010 Bank entered 43 villages offered loans, not other 32
- 2011-2012 We resurveyed all villages

# Kenneth Arrow 1999



“This leads to an important and long-standing question: does the market (or, for that matter, the large, efficient, bureaucratic state) destroy social links that have positive implications for efficiency?”

# Kenneth Arrow 1999



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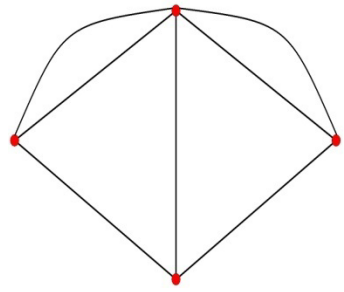
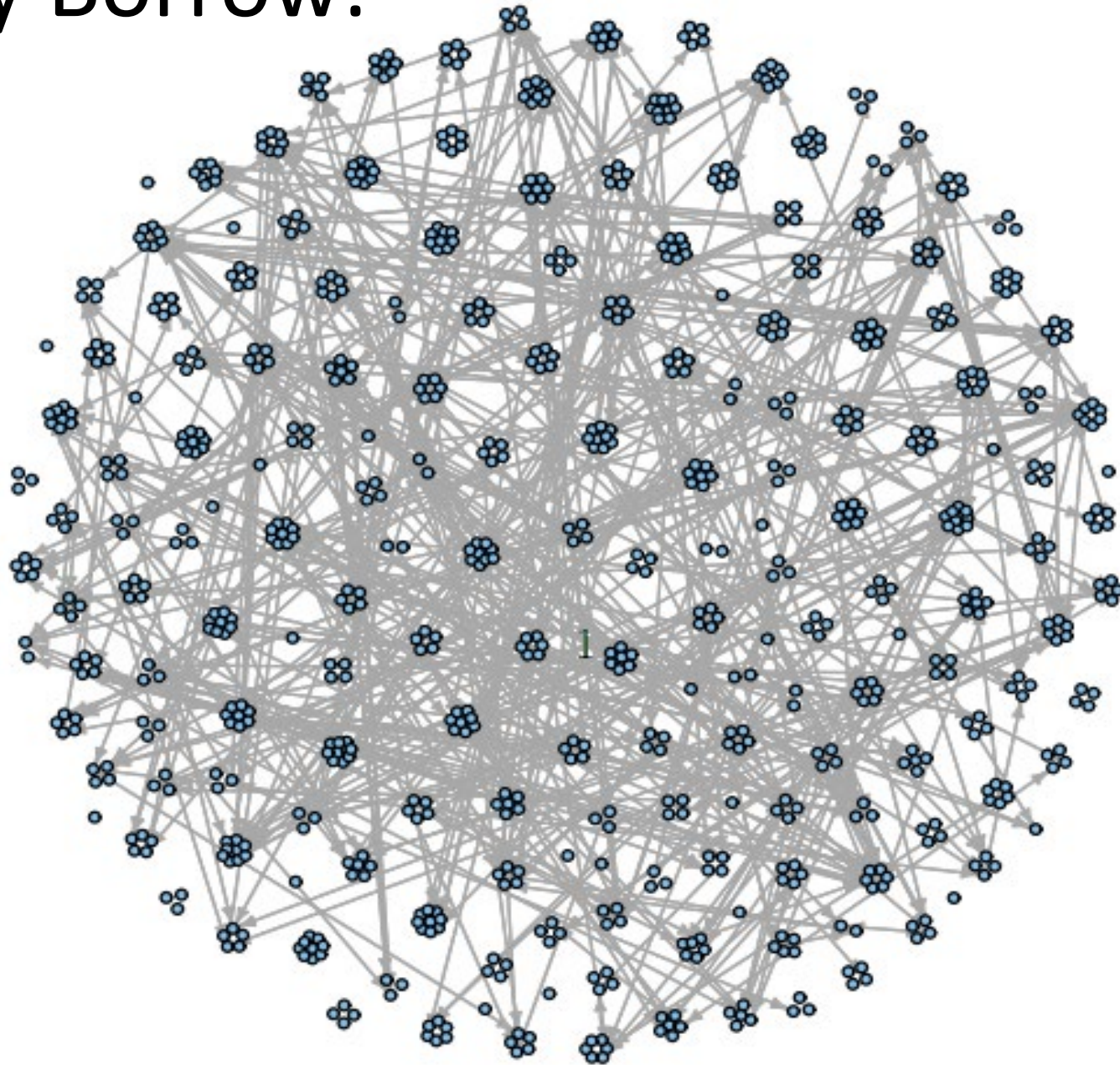
Here:

Does availability of formal loans change informal networks?

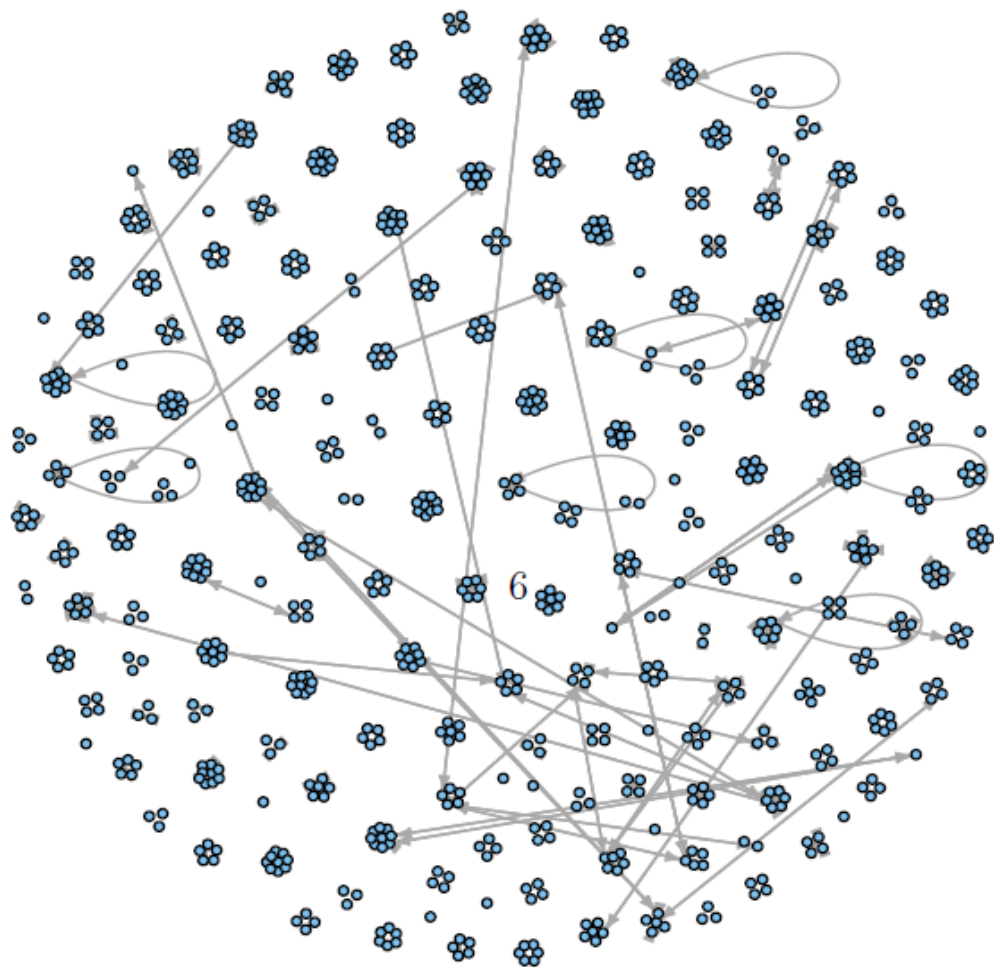
Which networks?



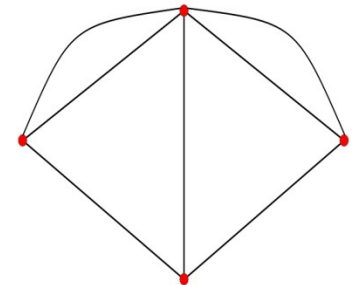
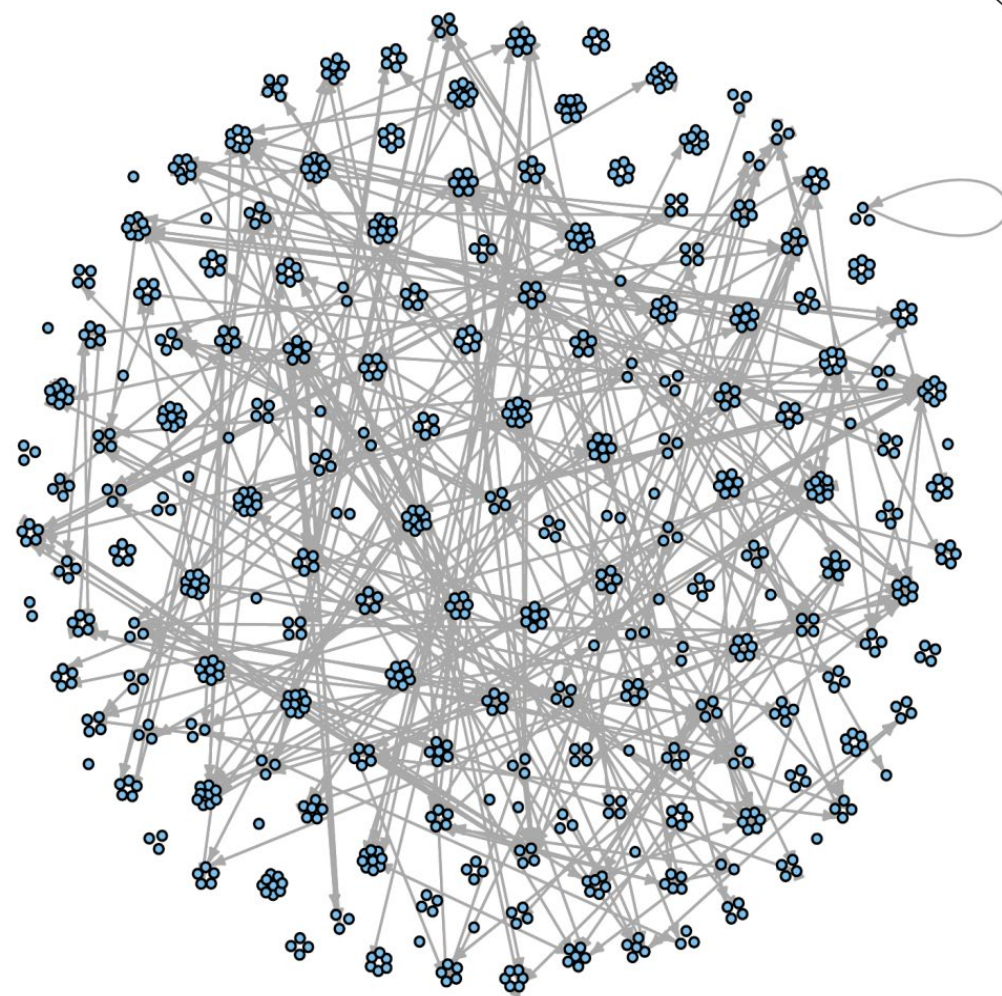
# Money Borrow:



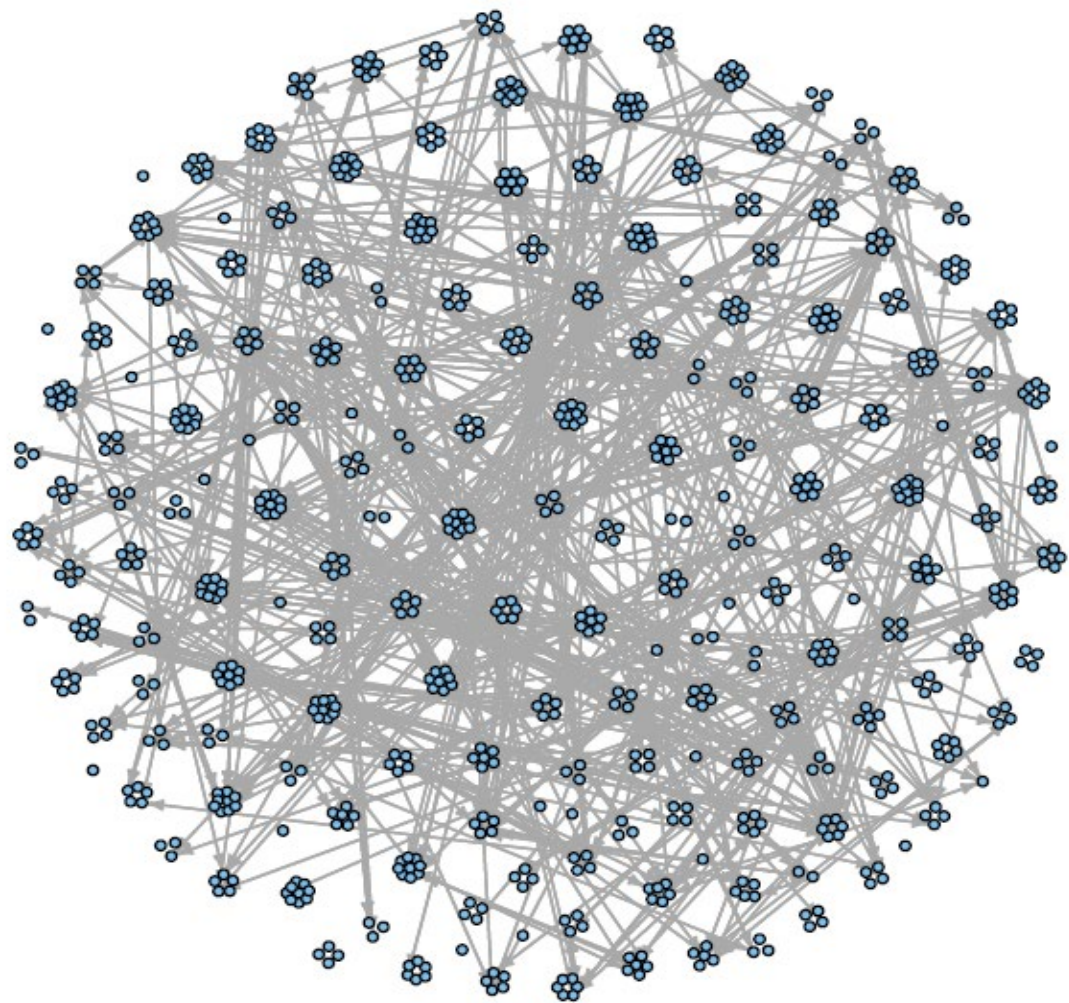
Temple



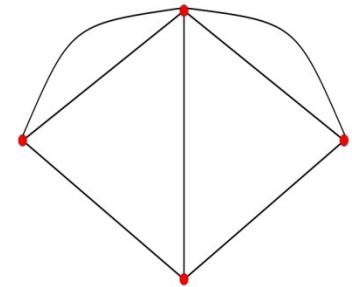
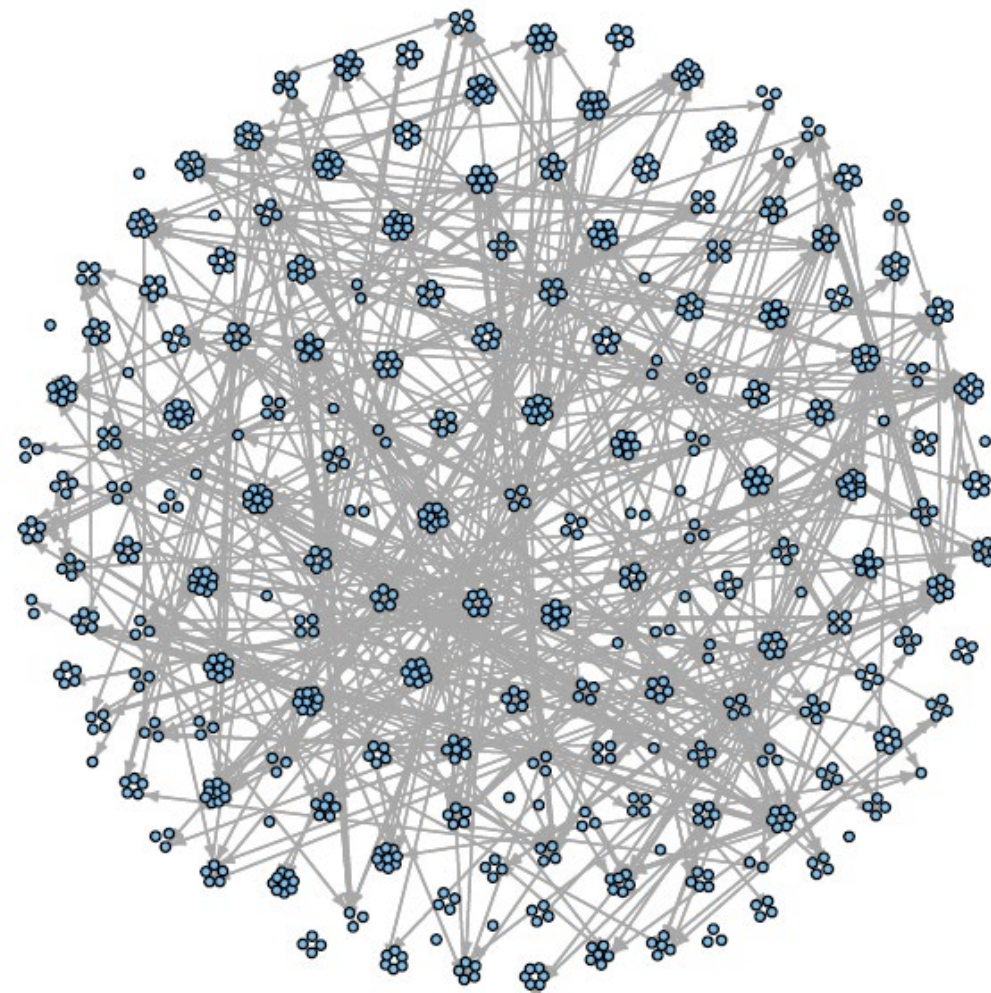
Advice



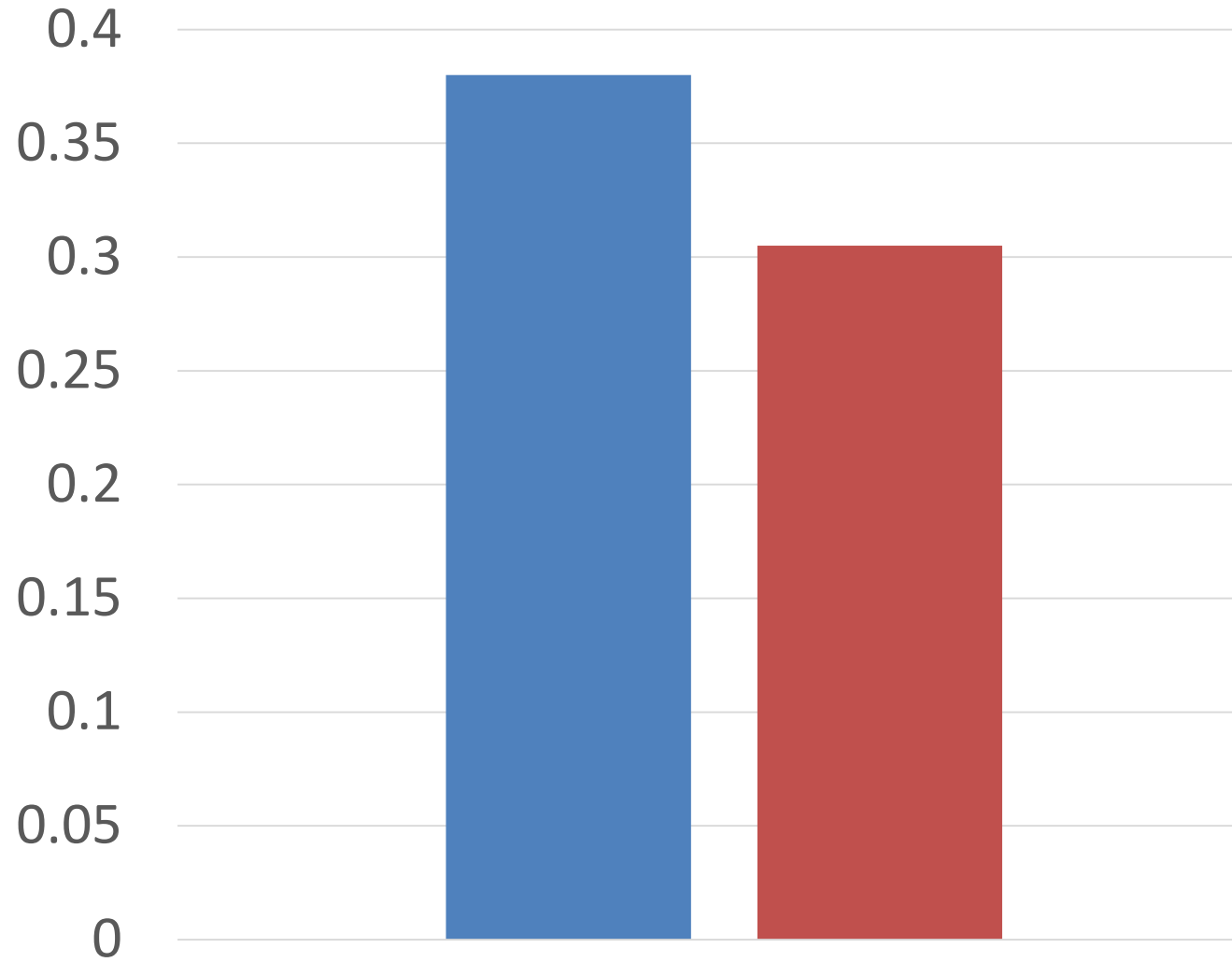
**Kero-Come**



**Medic**



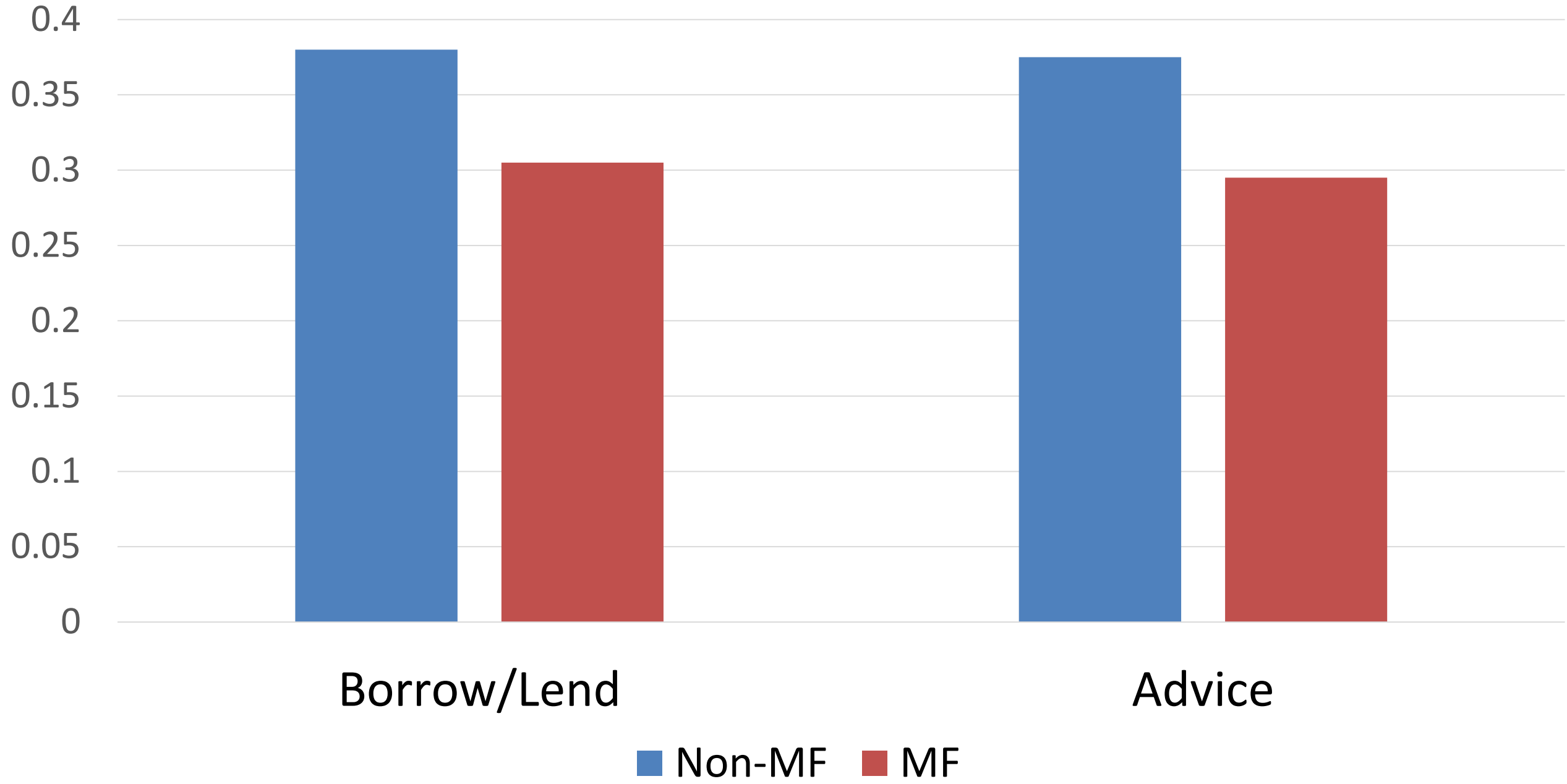
# Fraction of Relationships Retained 2012-2006



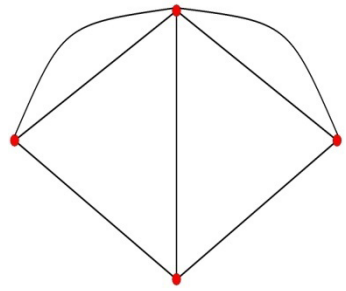
Borrow/Lend

■ Non-MF ■ MF

# Fraction of Relationships Retained 2012-2006

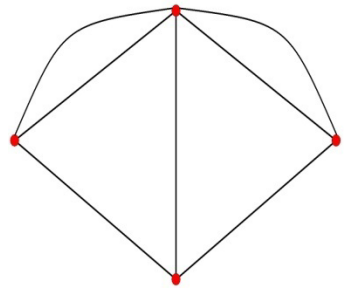


# Network Changes: Theory



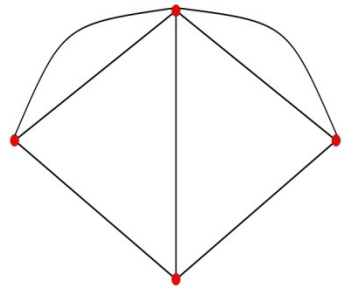
- BBCDJK 2024: Model
  - Takes time/effort to socialize to form/maintain relations
  - Loan participants decrease socializing
  - Then so do non-participants...
- Socializing changes *all* relationships, not just borrow/lending
- Multiplexing! What are its consequences?

# Outline



- Brief empirical look at multiplexing
- Do multiple layers affect diffusion (in an RCT)?
- Theory on how multiplexing impacts diffusion (simple, complex)
- Demographics and multiplexing

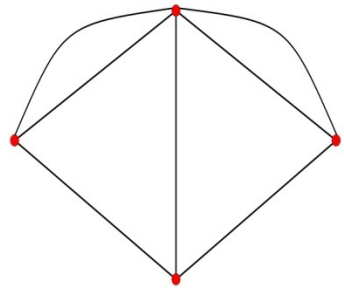
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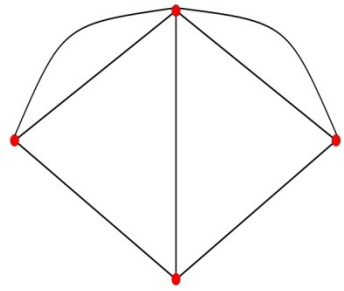


# Two Data Sets



- 75 Microfinance villages from BCDJ 2013,19, BBCDJK 2024
  - 9 layers of networks
- 68 different villages (also Karnataka) from RCT of diffusion BCDJ 2019
  - 5 layers of networks

# Networks:

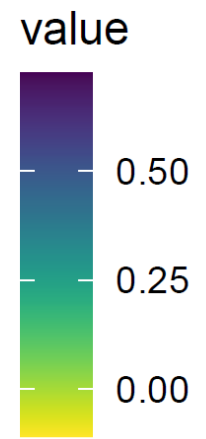
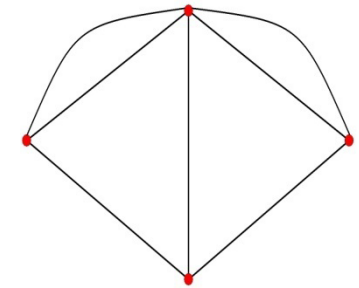
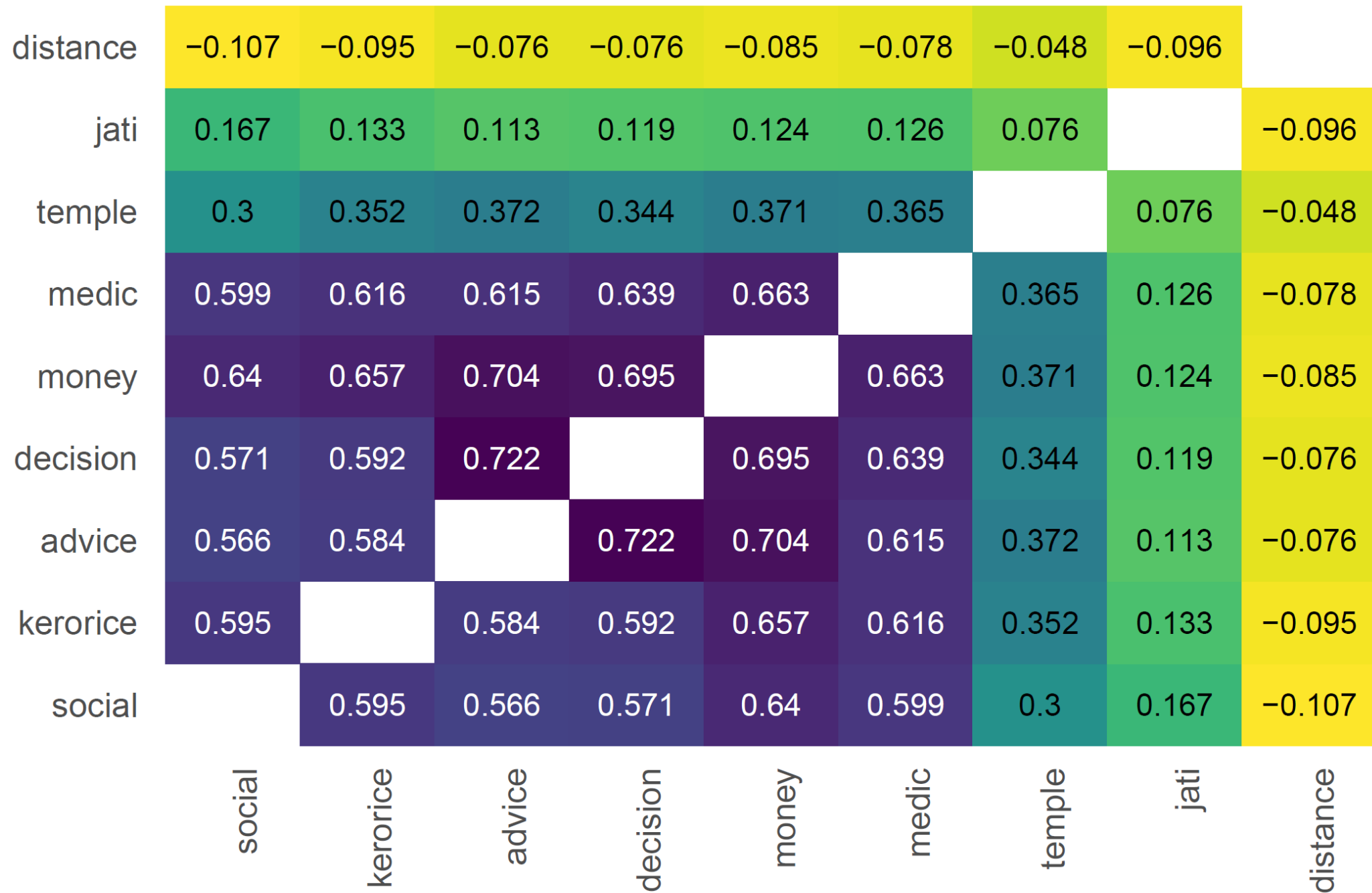


- MF villages 9 total:

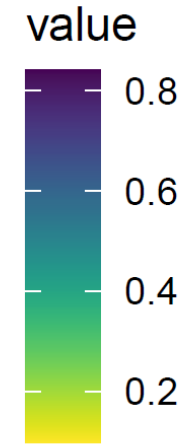
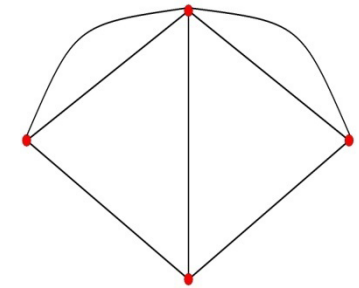
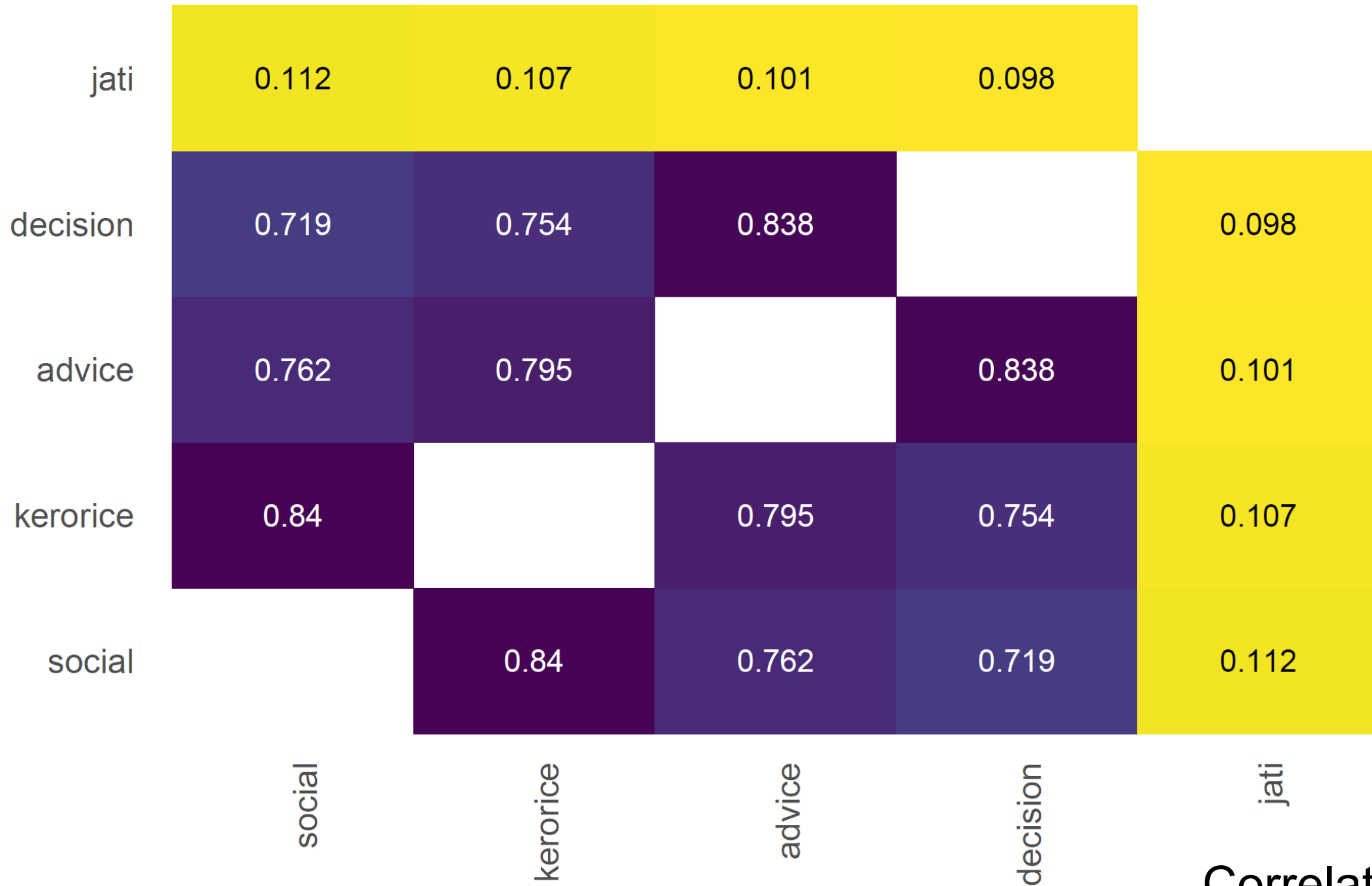
- Kero-Rice
- Money
- Socialize: relatives, non, visit
- Medical help
- Temple
- Advice
- Decision Help
- Jati
- Geography

- RCT villages 5 total:

- Kero-Rice/Money
- Socialize
- Advice
- Decision help
- Jati

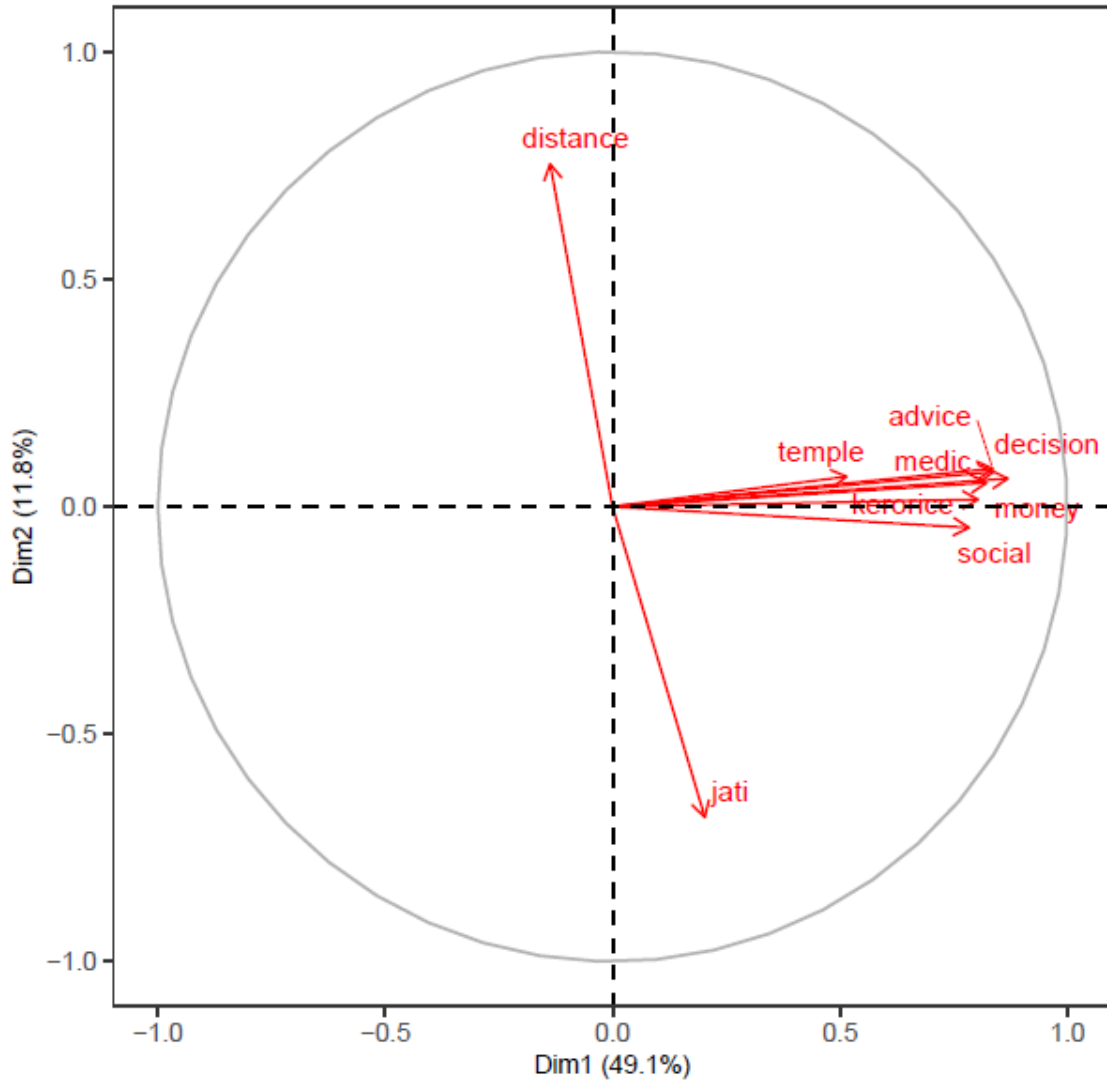


Correlations (MF)

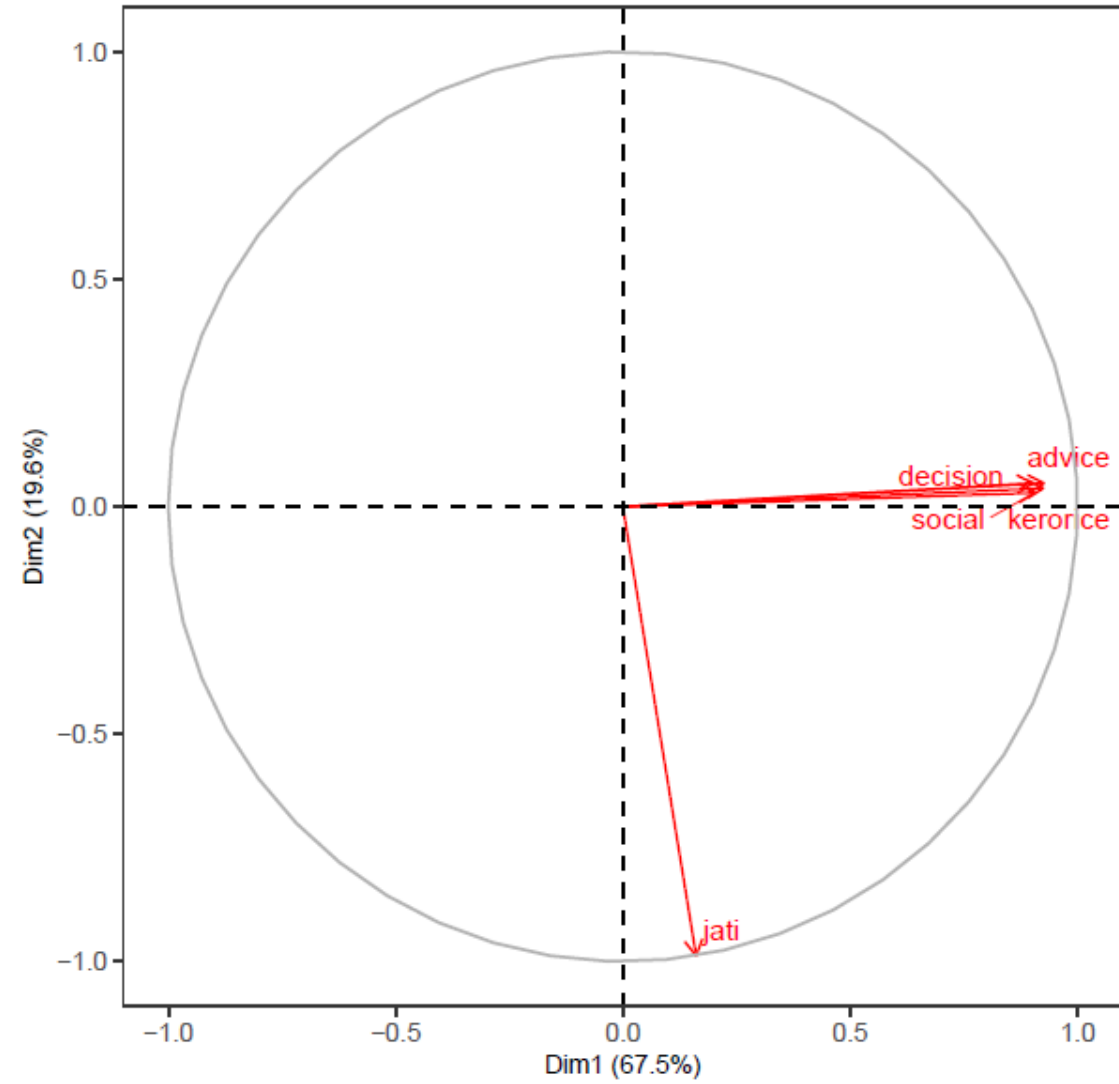


Correlations (RCT)

# Principal Component Analysis

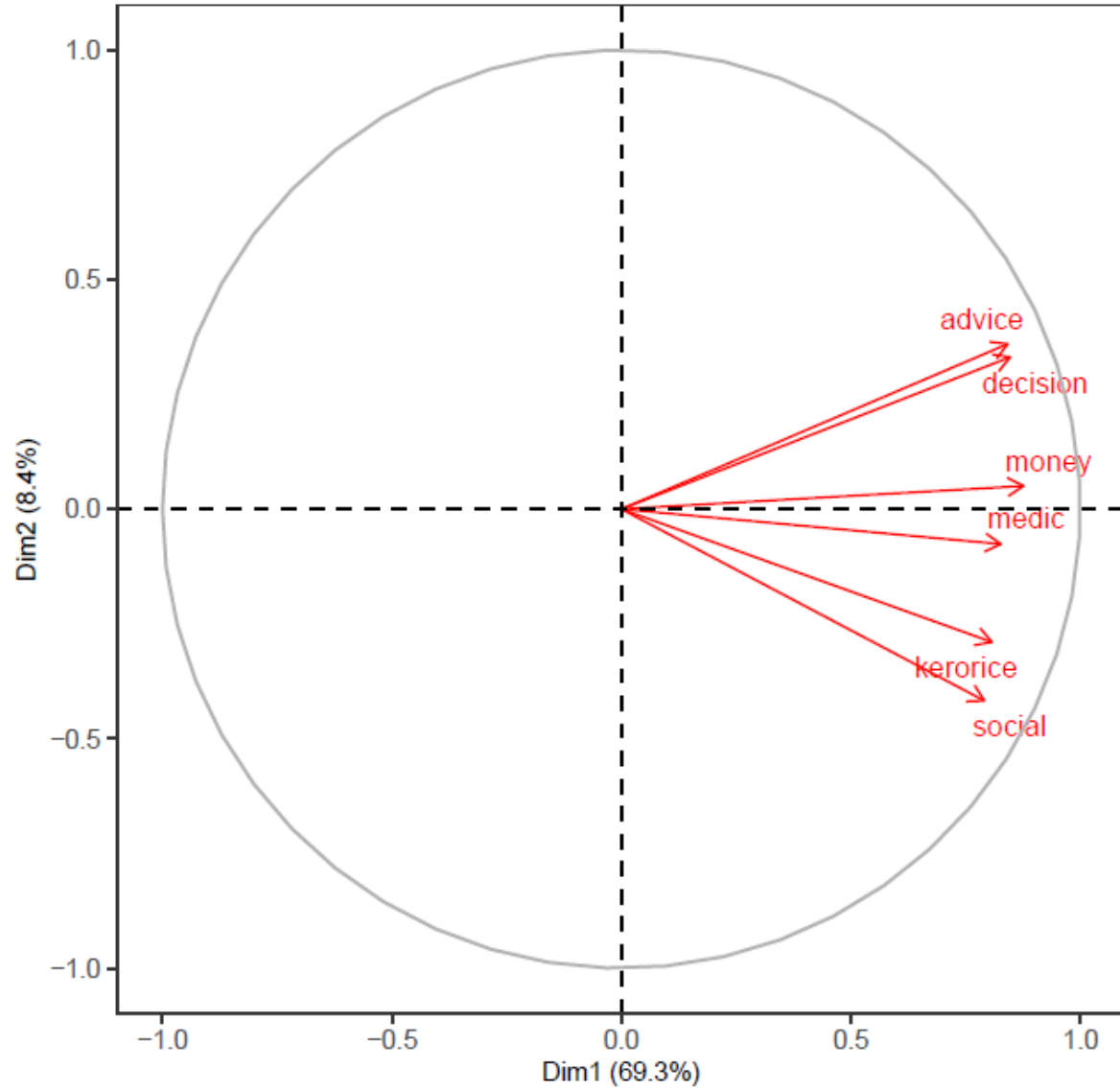


MF villages

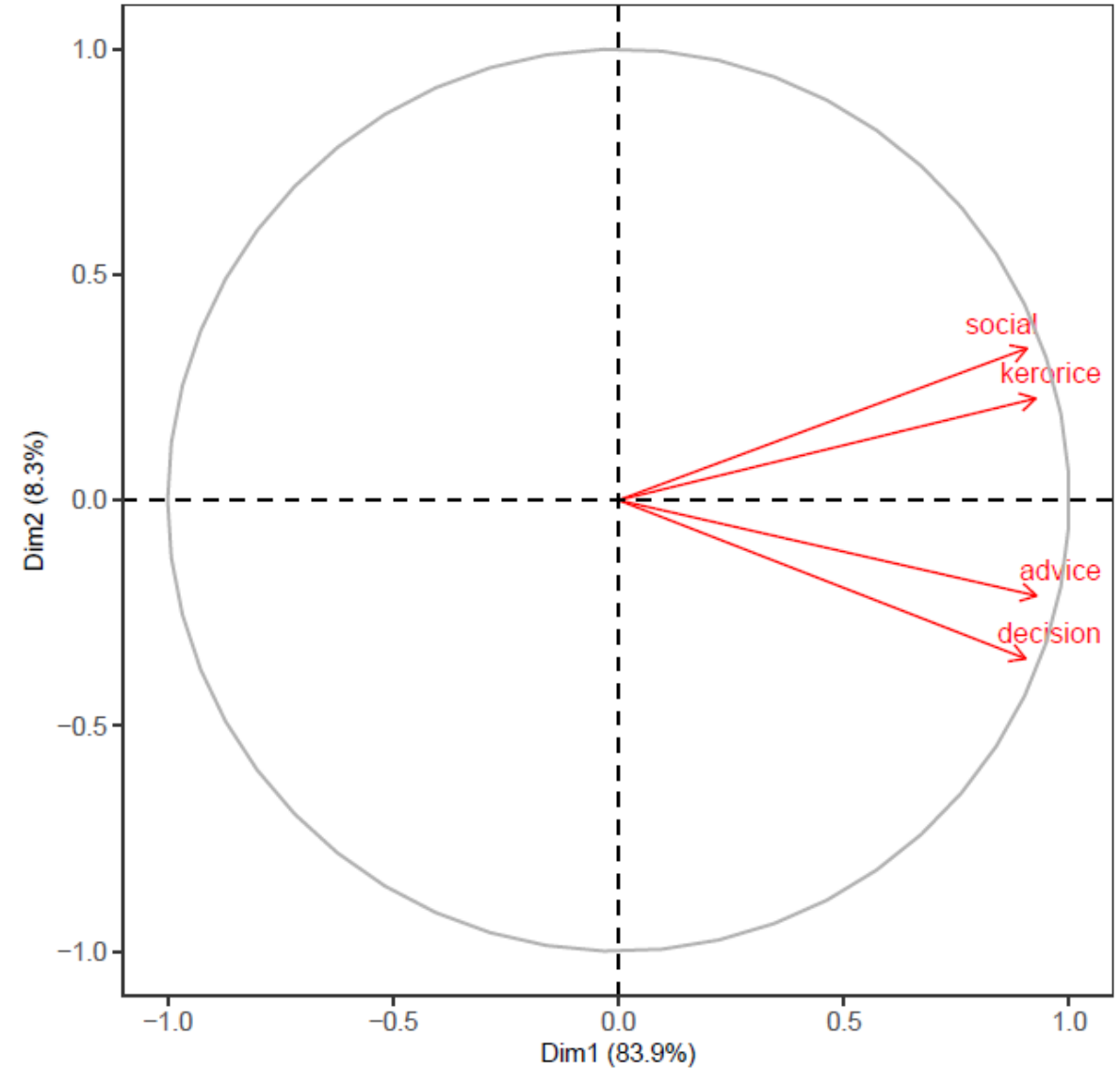


RCT villages

# Principal Component Analysis (excluding jati and geography)

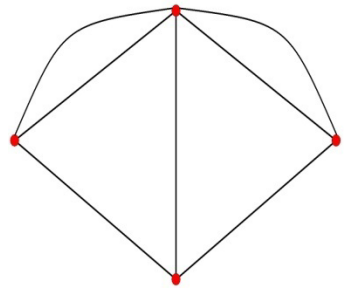


MF villages



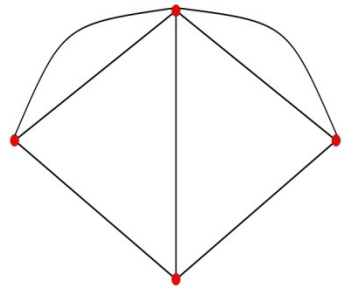
RCT villages

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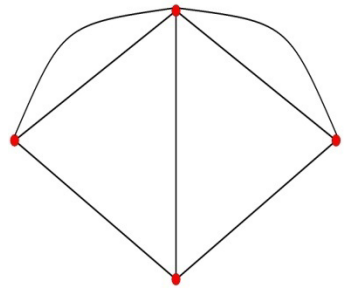
# RCT on Diffusion



- Spread information about a chance to win a cell phone
- Randomly tell 3 to 5 people in each village and ask them spread information
- Measure diffusion of information by how many people participate in the cell-phone giveaway

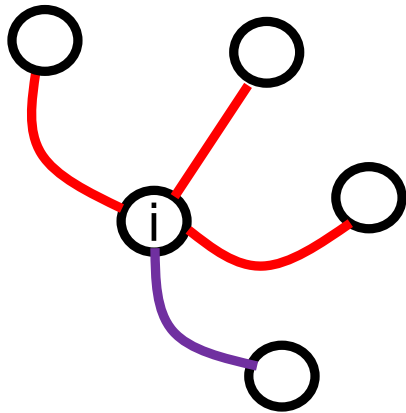


# Diffusion

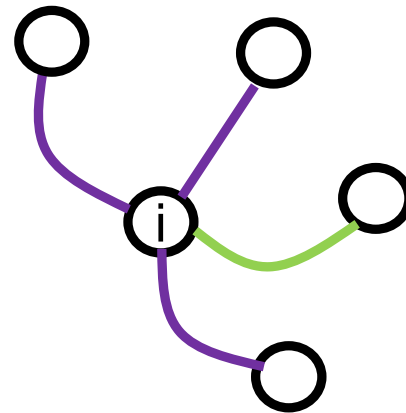


- Which network layer(s) best predict the diffusion?
  - Is it more important to have “seeds” central in advice or in kero/rice or in jati...?

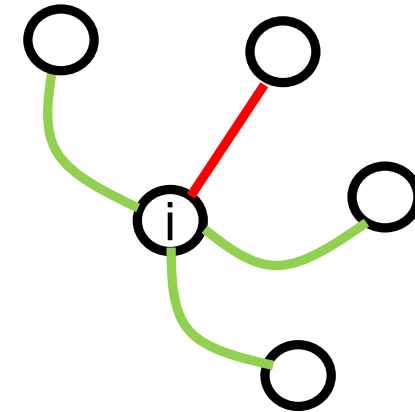
central in  
advice network



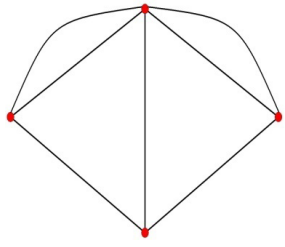
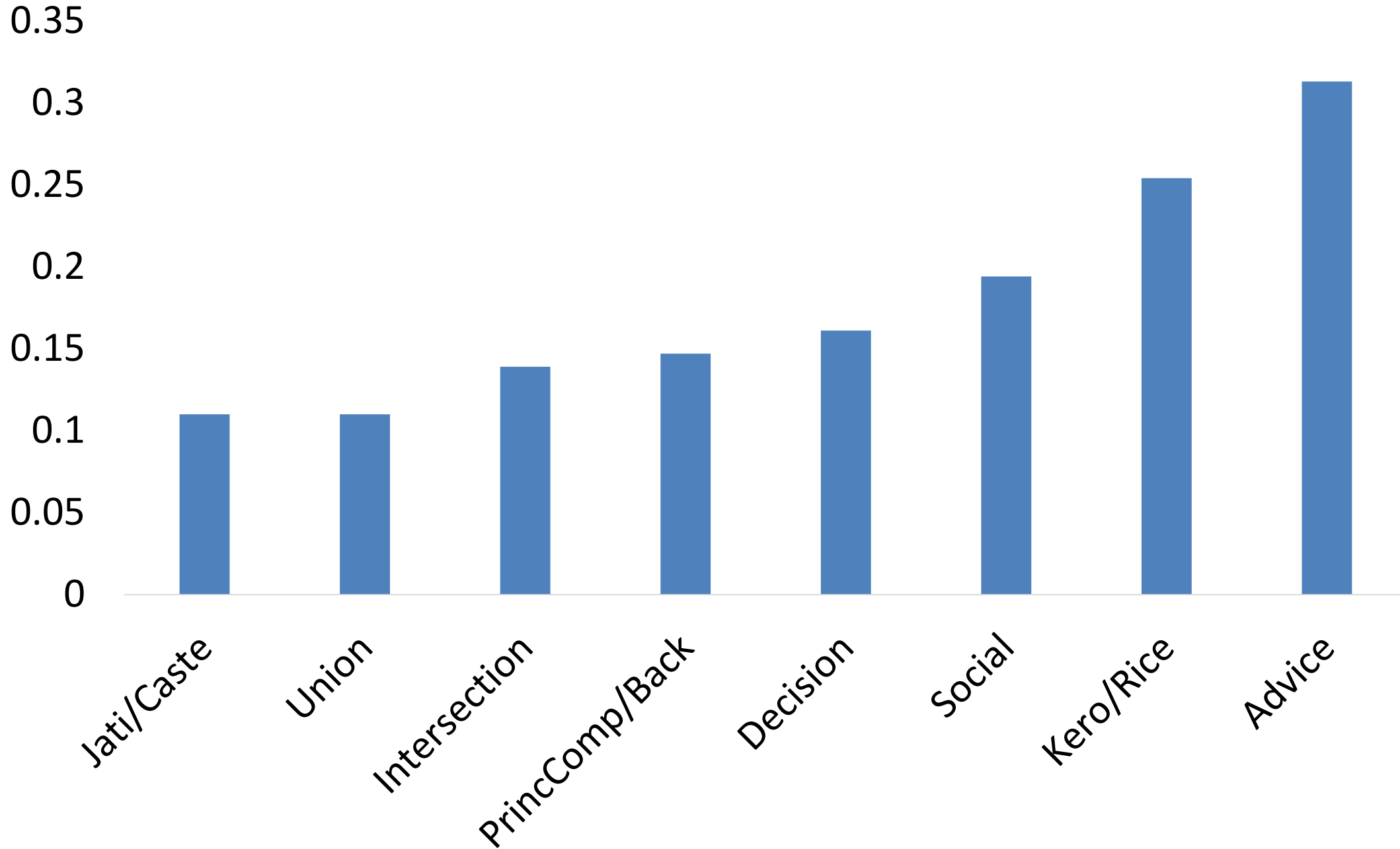
central in  
kero/rice network

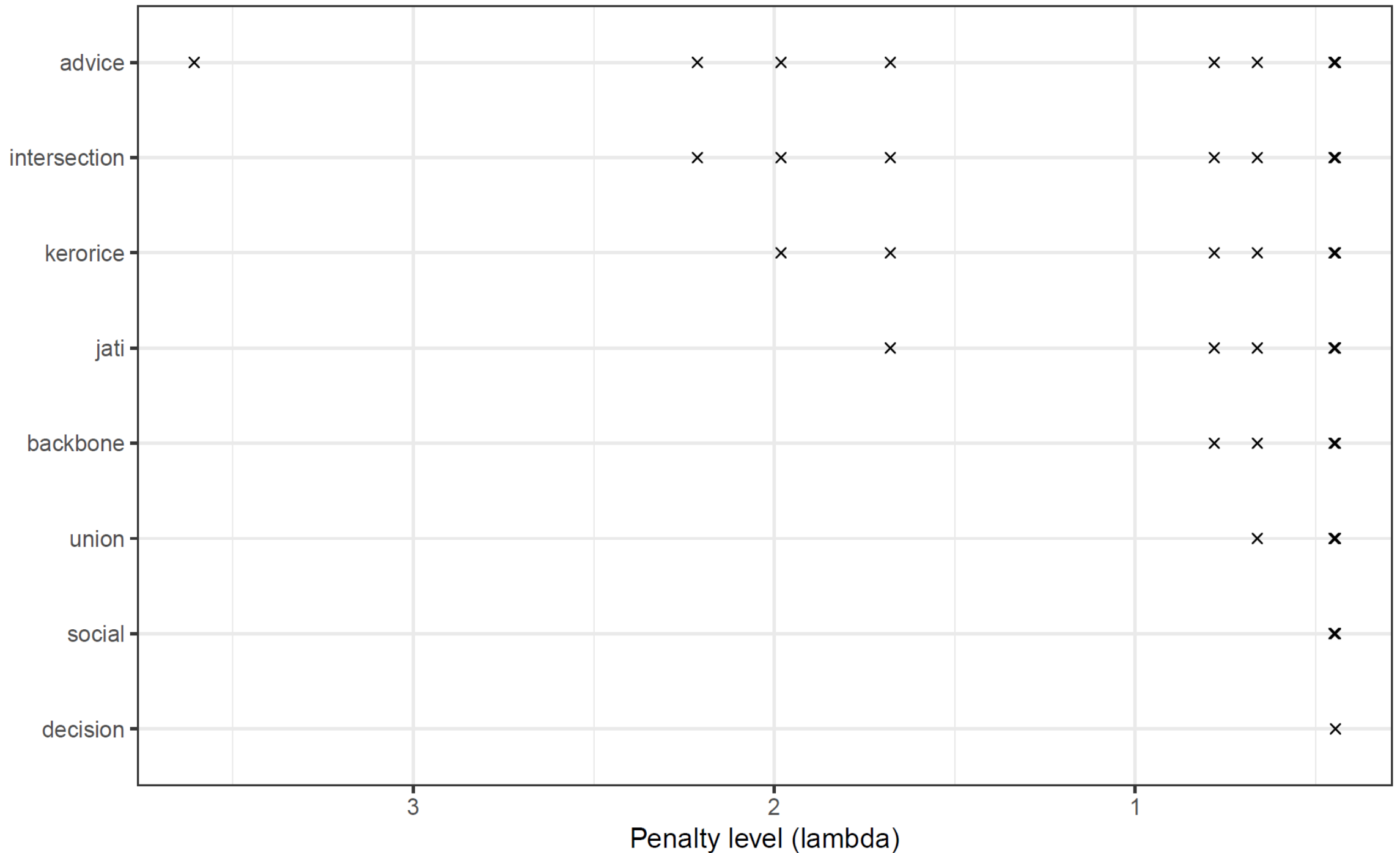


central in  
jati network

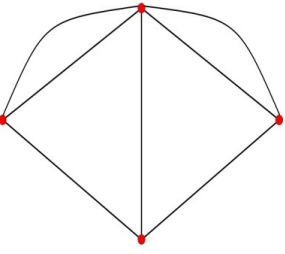


# Extent of Diffusion Explained (R-sq)





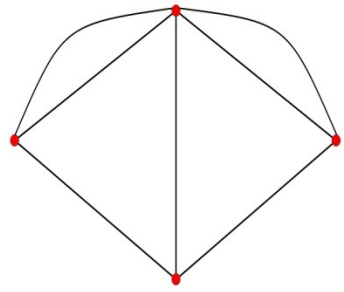
# Cumulative Predictive Power:



layer	df	R.sq.	F-stat	p-val
advice	1	0.233	20.057	0.000
intersection	2	0.276	3.888	0.053
kerorice	3	0.281	2.134	0.127
jati	4	0.325	2.844	0.045
backbone	5	0.336	2.415	0.058
union	6	0.340	1.971	0.096
social	7	0.342	1.657	0.147
decision	8	0.344	1.419	0.215

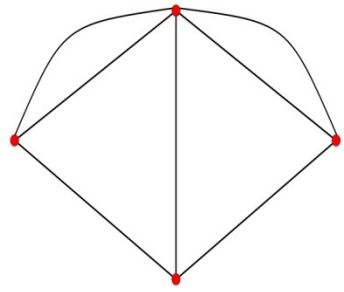
More than just advice matters:  
combinations of layers matter in predicting diffusion

# Which Networks?



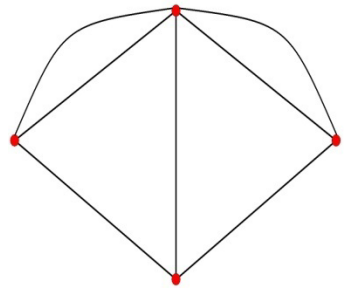
- Advice, kero/rice, social, are most predictive of diffusion
- Jati is not directly predictive of diffusion, but does significantly add when combined with advice and others...
- *Multiple layers matter, and matter differently*

# Outline



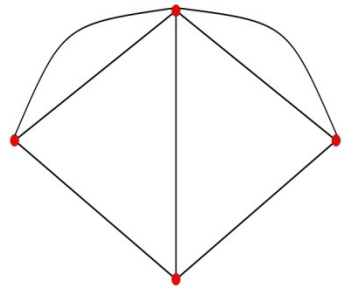
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# Theory: Diffusion



- Agents either infected or susceptible
- Transition back and forth in discrete periods
- If infected recover randomly at some rate  $\delta$  each period
- If susceptible can be infected by contact with infected neighbor

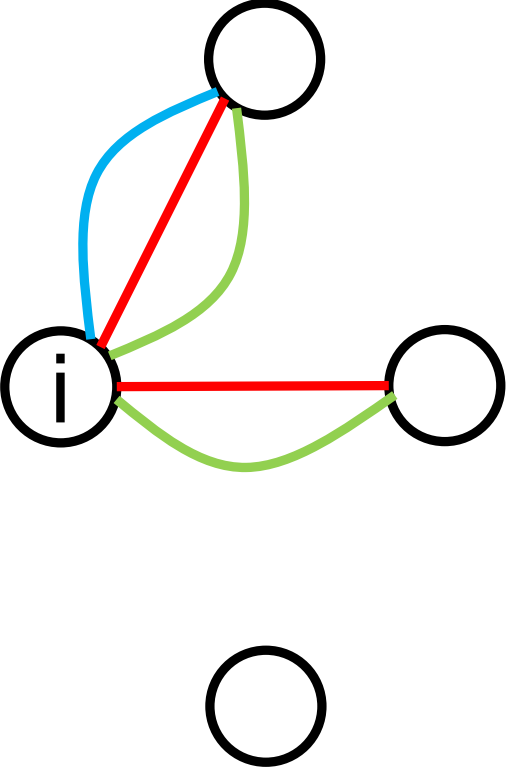
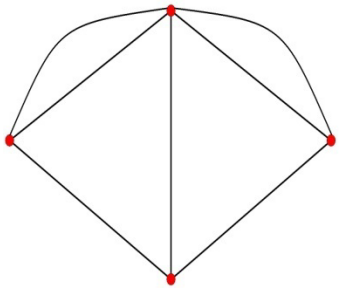
# Theory: Diffusion



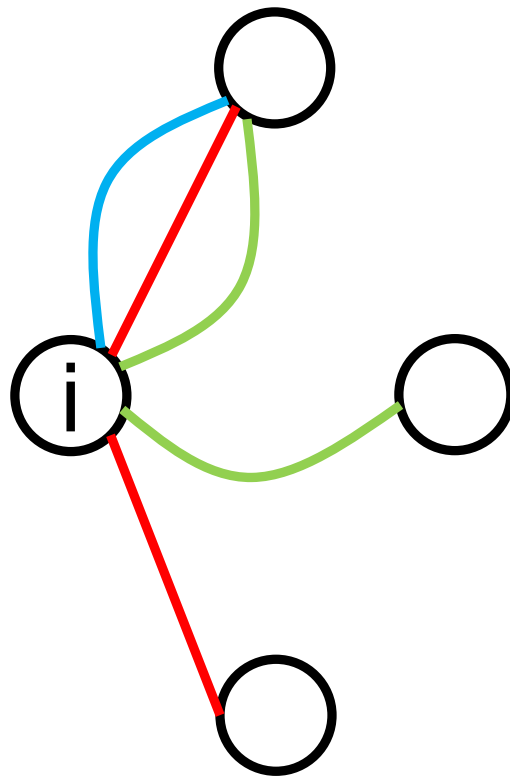
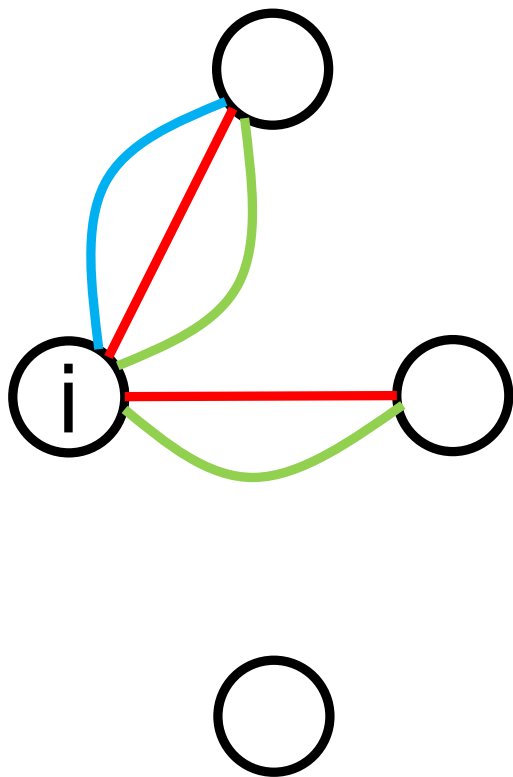
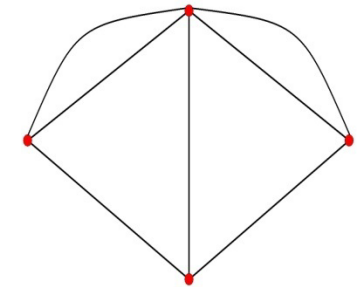
- Probability  $q^l$  that infected agent passes infection on layer  $l$
- Agent becomes infected when getting at least  $\tau$  'messages'
  - $\tau = 1$  simple contagion
  - $\tau > 1$  complex contagion
- Passing along layers independent (not too negatively correlated)



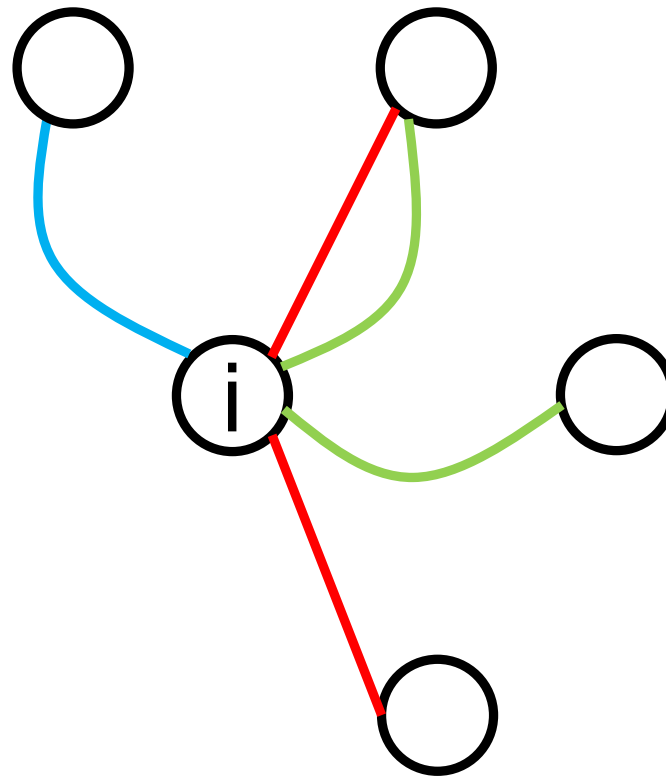
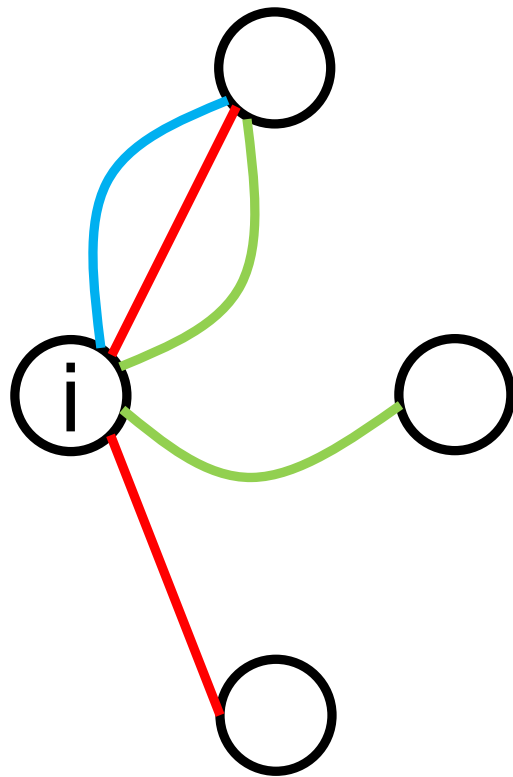
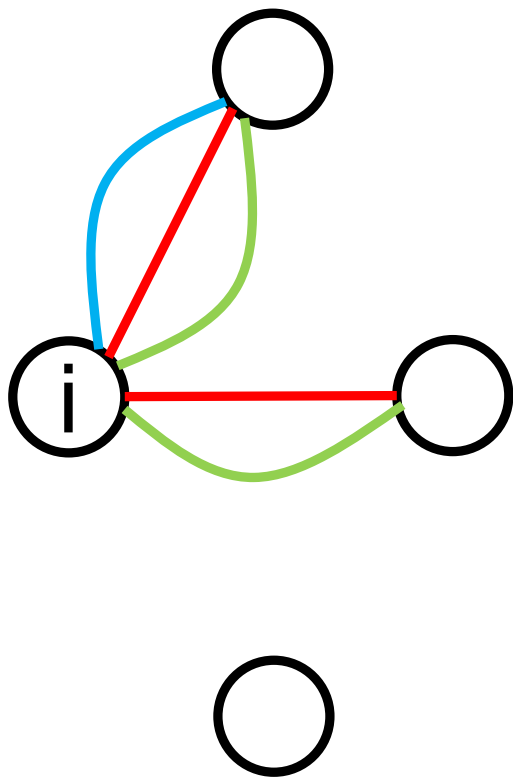
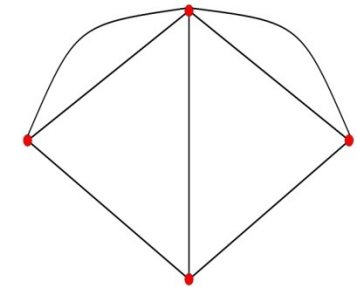
# Less Multiplexed



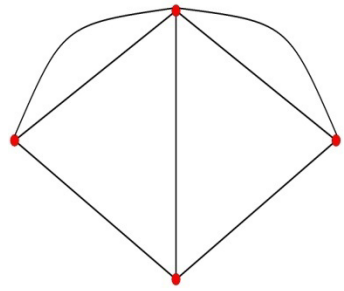
# Less Multiplexed



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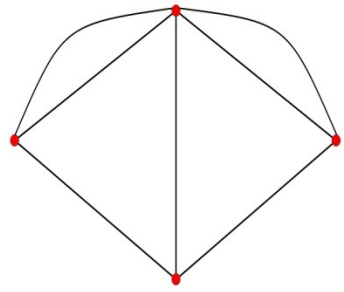
# Proposition: Multiplexing Hurts Diffusion under *Simple Diffusion*



Consider two agents  $i$  and  $j$ , with  $i$  is less multiplexed.

If  $i$ 's and  $j$ 's neighbors are each infected with probability  $p$  then  $i$  is more likely to be infected.

# Proposition: Multiplexing Hurts Diffusion under *Simple Diffusion*

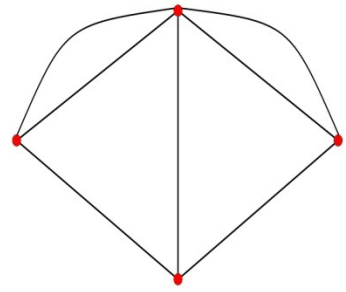


Consider agents  $i$  and  $j$ , with  $i$  *more* multiplexed than  $j$ .

If  $i$ 's and  $j$ 's neighbors are each infected with probability  $p$ , then  $i$  is *less* likely to be infected.

*Less multiplexing – more diffusion/contagion*

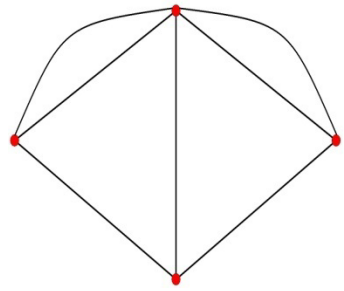
# Proposition: Multiplexing Hurts Diffusion under *Simple Contagion*



In an SIS (or SIR model), the mean-field steady-state infection rate is **decreasing** in the multiplexing of the network.

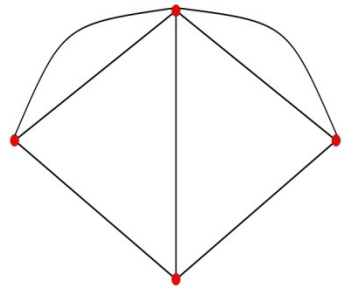
*Less multiplexing - more diffusion/contagion*

## Intuition/Proof Logic:



- $p$  chance neighbor is infected
- Infection on one multiplexed relation:  $q^A p + q^B p - q^A q^B p$
- Infection on two un-multiplexed relations:  $q^A p + q^B p - q^A q^B p^2$

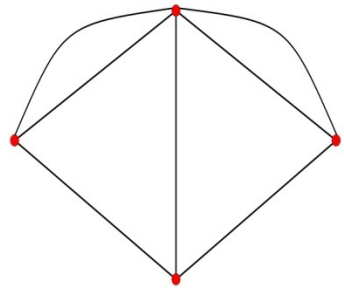
## Intuition/Proof Logic:



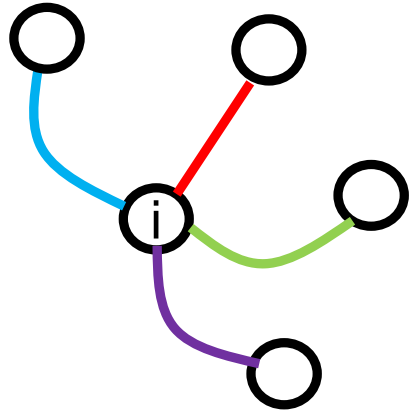
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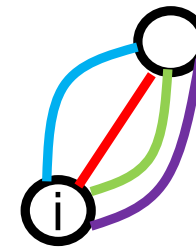
# Back to the Data



Theory: more multiplexed networks less simple diffusion/contagion



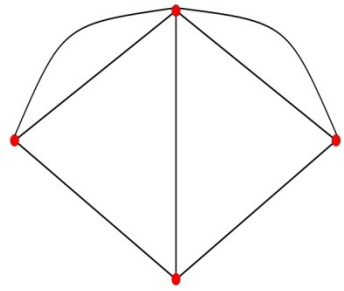
more diffusion



less diffusion

Do we see less diffusion in more multiplexed villages?

# Multiplexing Index

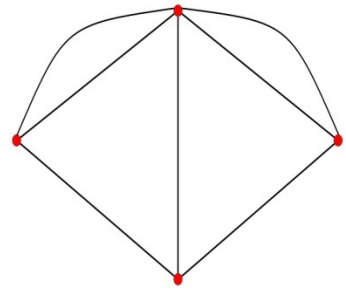


$$m_v := \frac{1}{n} \sum_i \frac{\sum_j \left( \sum_l g_{ij,v}^l / L \right)}{\sum_j \mathbf{1}\{\sum_l g_{ij,v}^l > 0\}}$$

	Participation
High Multiplexing x Seed Set Centrality	-.039**
	(.017)
Seed Set Centrality	.052***
	(.016)
High Multiplexing	-.023
	(.016)
Observations	68

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High Multiplexing x Seed Set Centrality	<b>-.039**</b>
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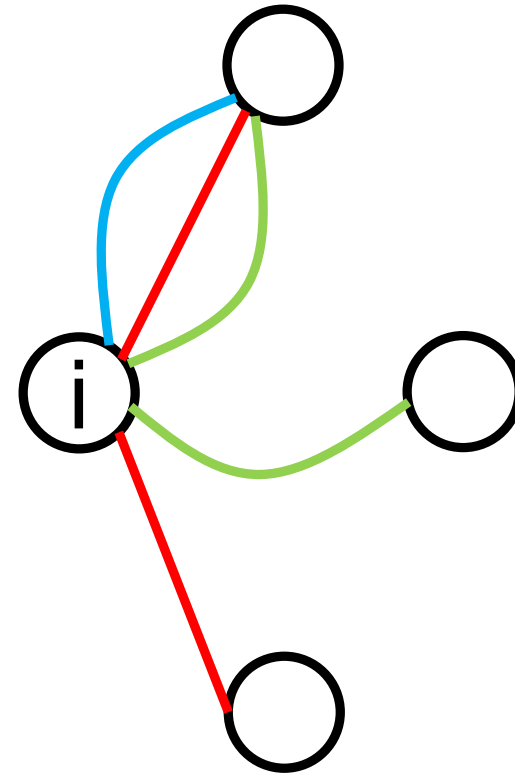
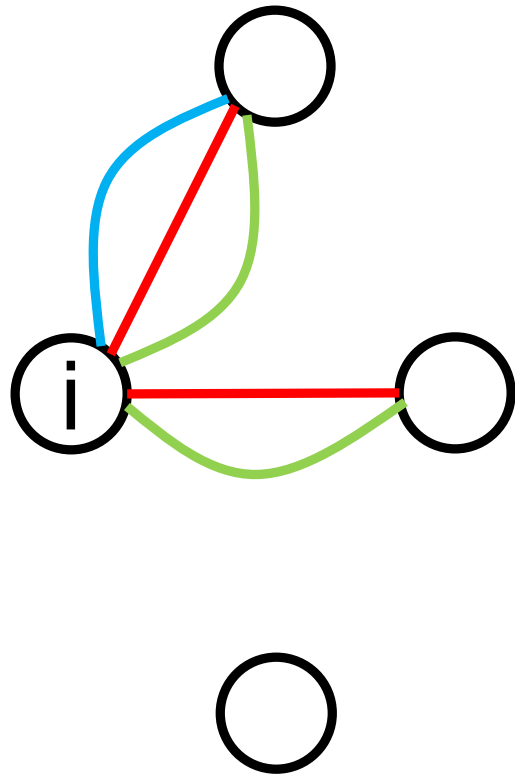
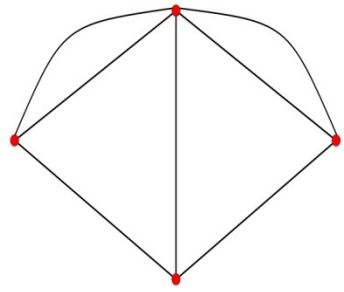
# Complex Diffusion is more Complex



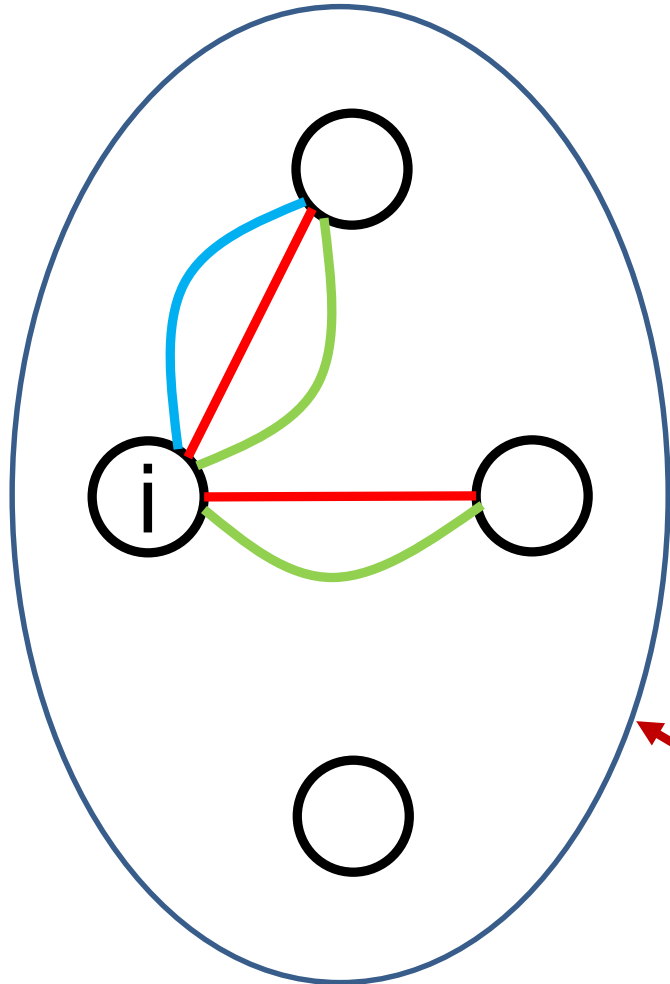
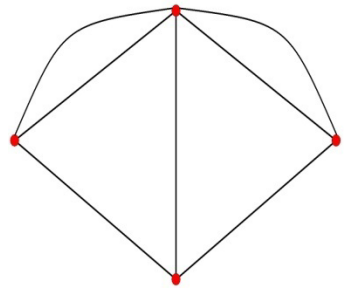
No longer can order based on multiplexing

Interaction with threshold

# Complex Contagion

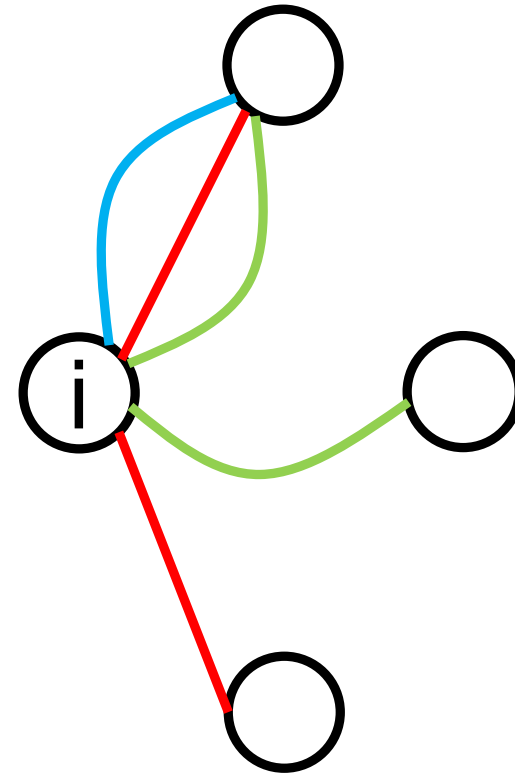


# Complex Contagion

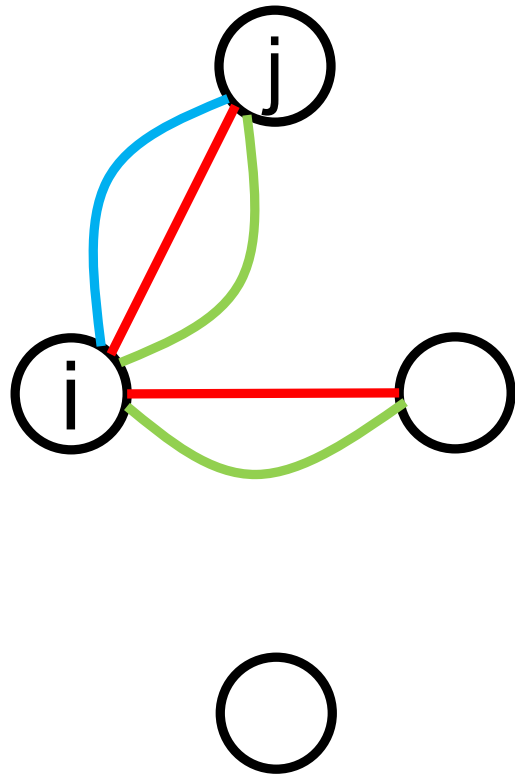
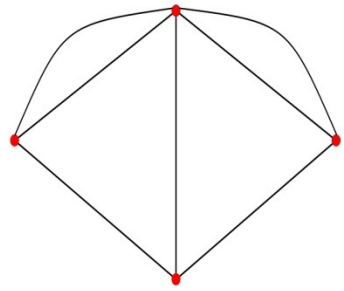


$\tau = 5$   
need all  
neighbors

Easier to infect

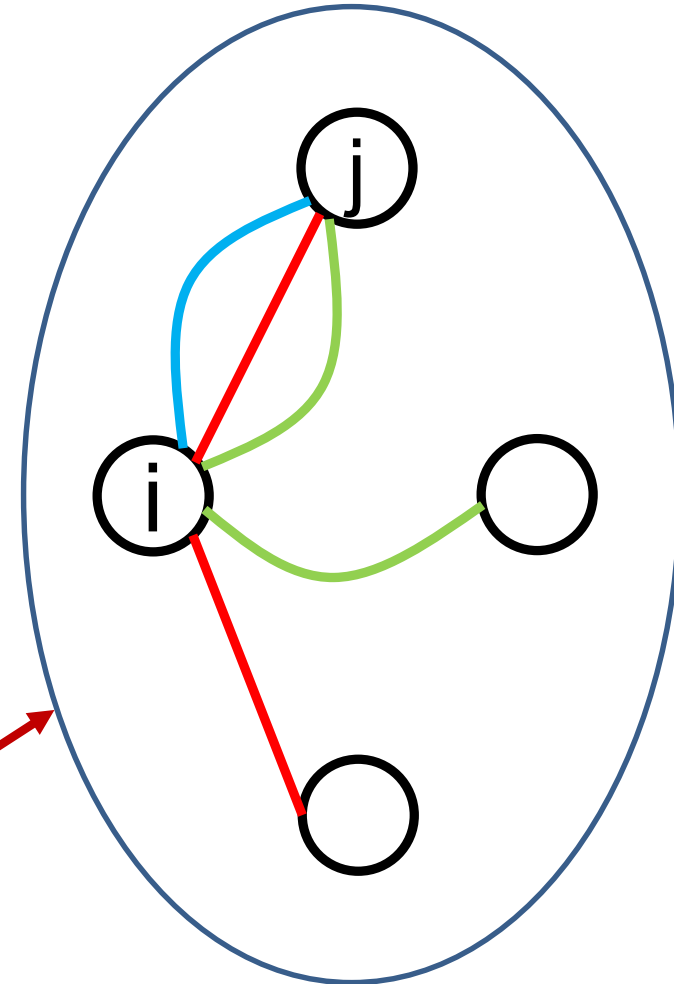


# Complex Contagion



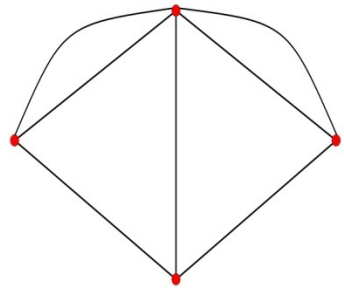
$\tau = 4$   
need  $j$  and  
one  
other

Easier to infect





# Complex Contagion

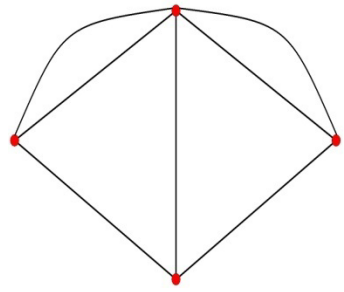


*Competing forces*

Less multiplexing – more chances to reach infected individuals

More multiplexing - more chances to get reinforced/multiple activations

# Proposition: Complex Contagion

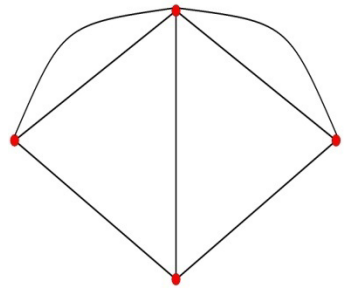


With **complex** contagion (threshold  $>1$ ), total degree summed across layers is at least  $\tau + 1$  :

There exist  $p' < p''$  (increasing in threshold) such that

- for  $p < p'$  infection probability is **increasing** in multiplexing;
- for  $p > p''$  infection probability is **decreasing** in multiplexing.

# Complex Contagion



When the infection rate is low,

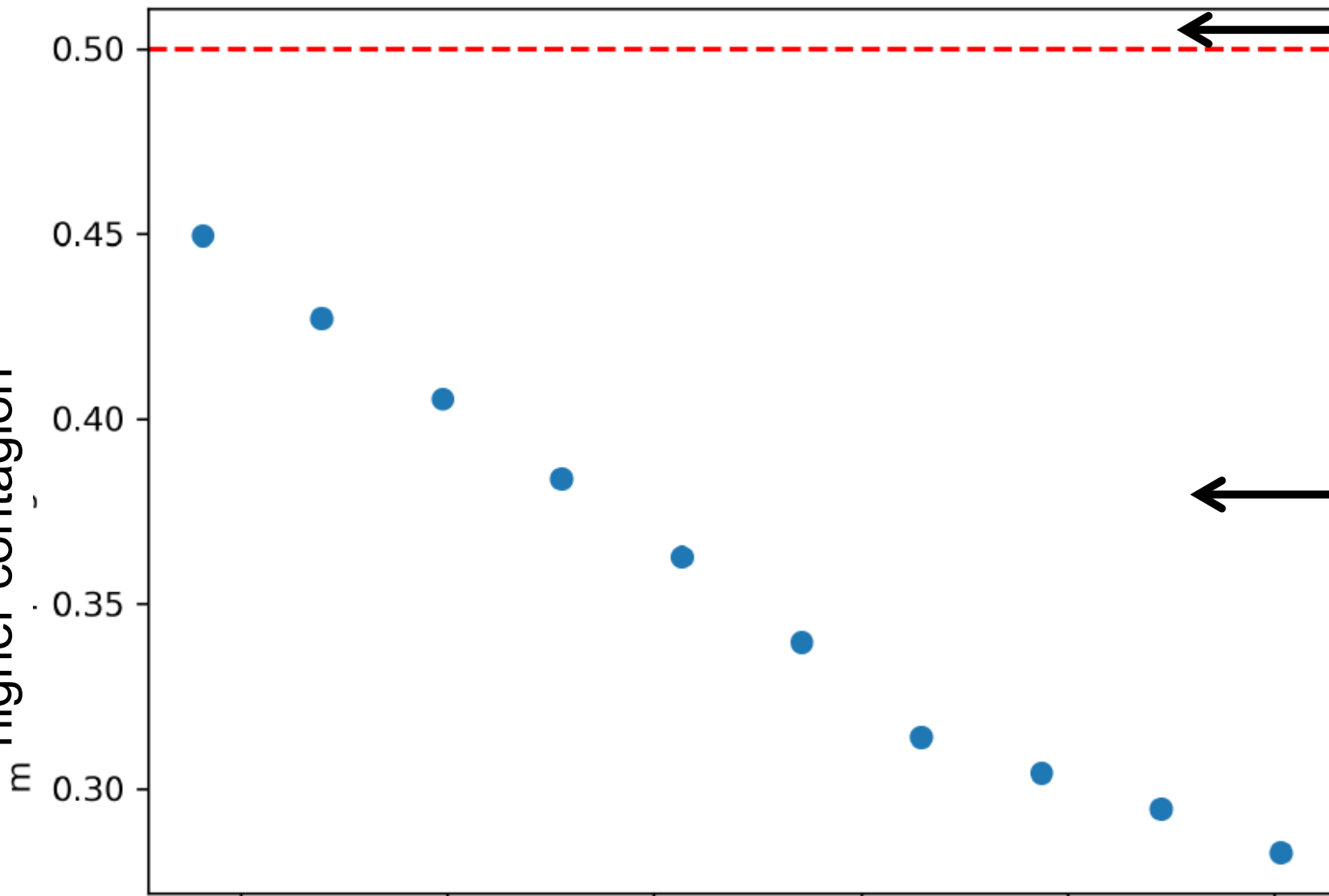
- Two neighbors being infected is unlikely.
- A multiplexed neighbor infected gives twice contact.

When the infection rate is higher,

- More likely that only need one contact from these two links
- Non-multiplexed more likely to get at least one contact.

# Simulations on Villages, Simple ( $\tau = 1$ )

Fraction Sims that More  
Multiplexed network has  
higher contagion

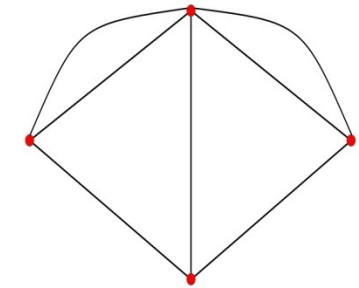


Multiplexing  
increases  
diffusion

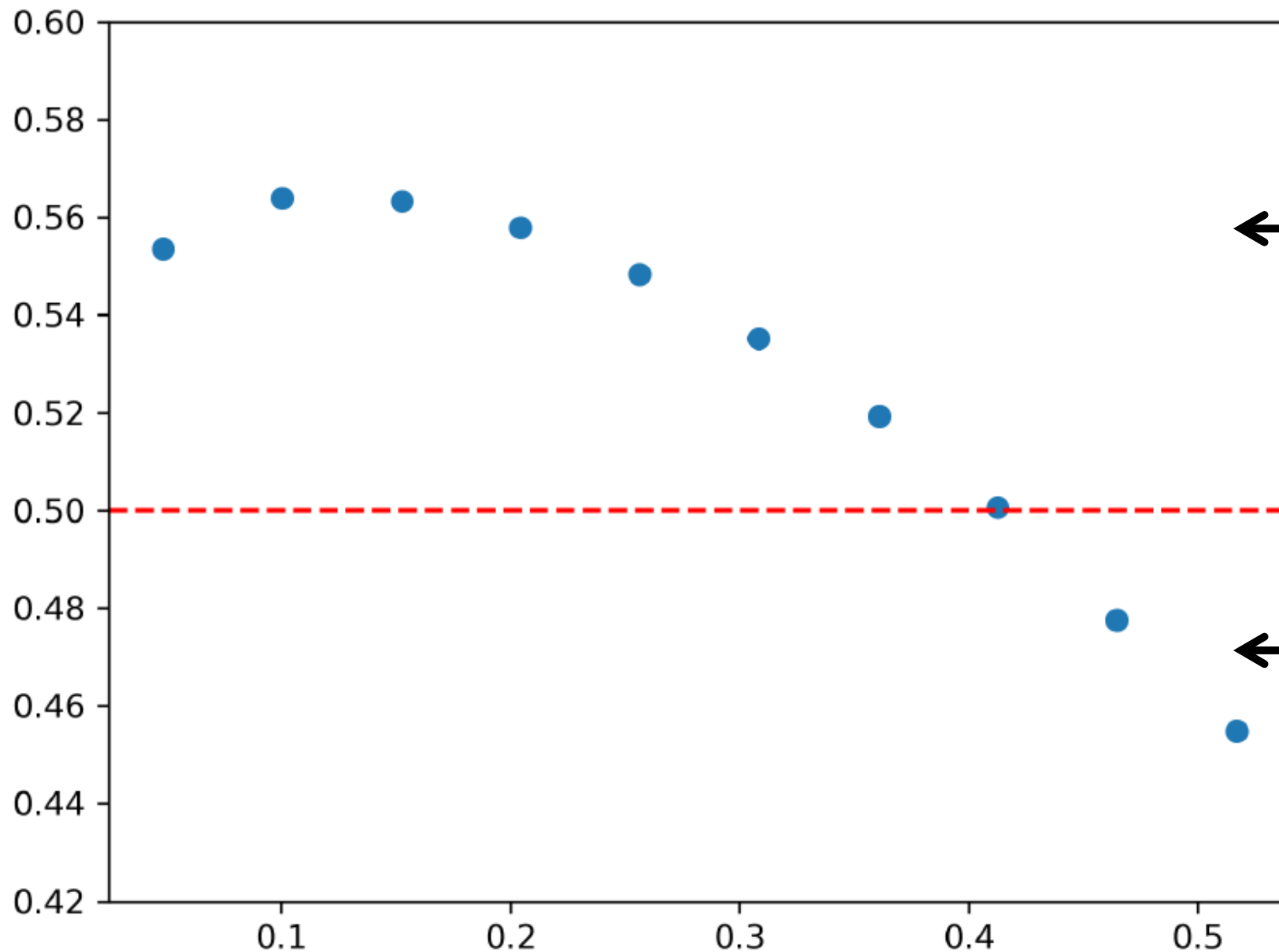
Multiplexing  
decreases  
diffusion

Contagion Level

# Simulations on Villages, Complex ( $\tau = 2$ )



Fraction Sims that More  
Multiplexed network has  
higher contagion

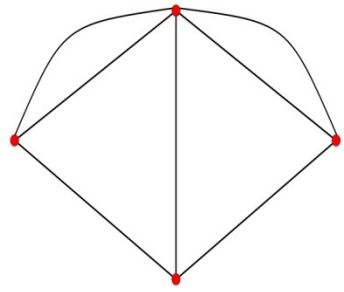


Multiplexing  
increases  
diffusion

Multiplexing  
decreases  
diffusion

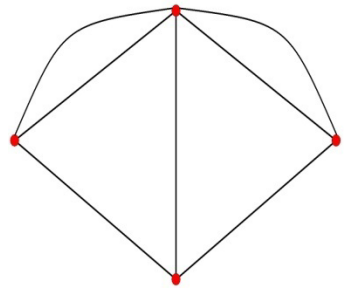
Contagion Level

# Outline

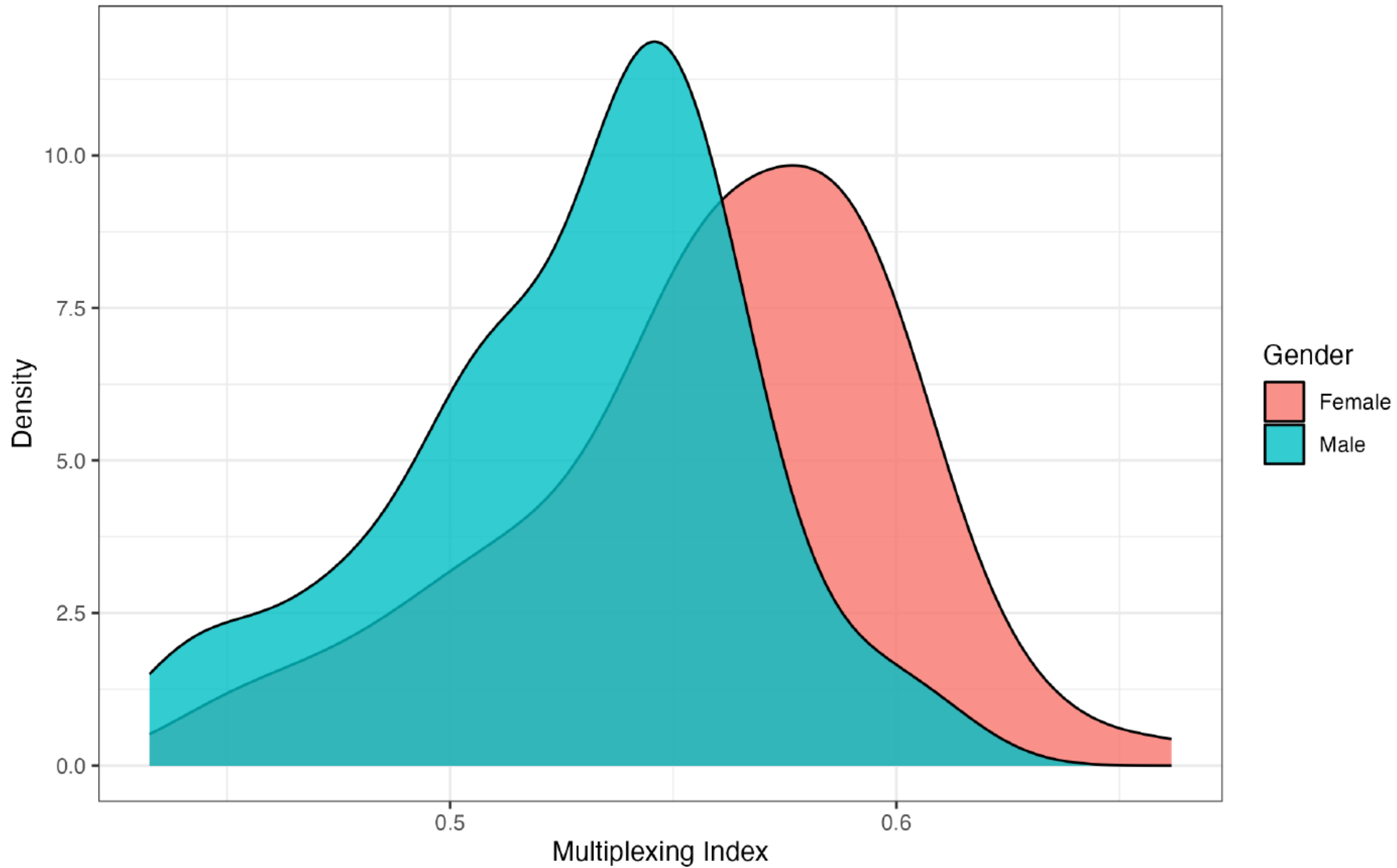


- Brief empirical look at multiplexing
- Do multiple layers affect diffusion (in an RCT)?
- Theory on how multiplexing impacts diffusion (simple, complex)
- **Demographics and multiplexing**

# Who Multiplexes?

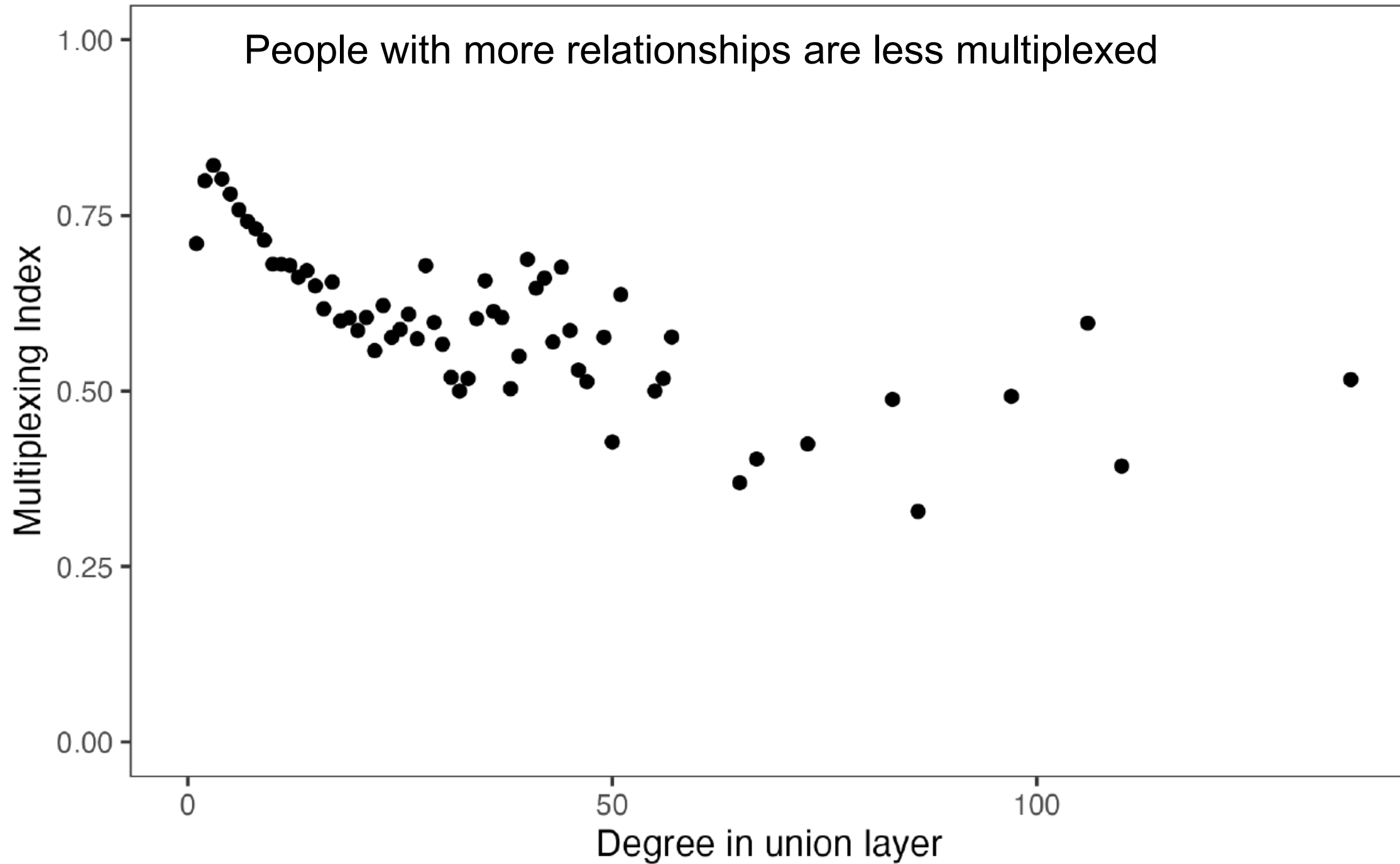


- Does multiplexing vary across individuals?
- Does multiplexing vary across locations?
- Which ones are more multiplexed?

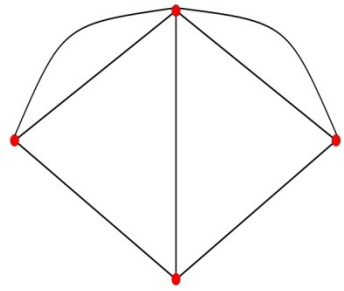




People with more relationships are less multiplexed

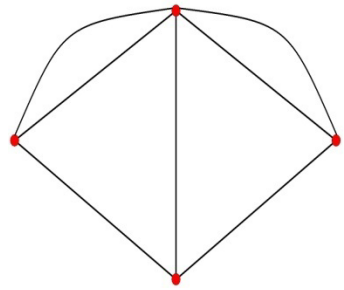


# Summary



- High multiplexing in data: high correlation between layers
- Some layers more predictive of diffusion, but multiple layers matter
- Multiplexing inhibits simple diffusion in a model and RCT, can aid complex
- Multiplexing higher among the less connected, females

# Closing Thoughts



- Multiplexing matters in many contexts
  - Networks between nations (trade, migration, war)
  - Networks between companies (partnerships, lending, competition)
  - Networks among workers (communication, direction, collaboration)
  - Networks among students (friends, study partners, roommates)
  - ....
- Need for more theory/empirics of multiplexing and behaviors, network formation, methods