

Social Capital in Networks

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Data Mining Lab

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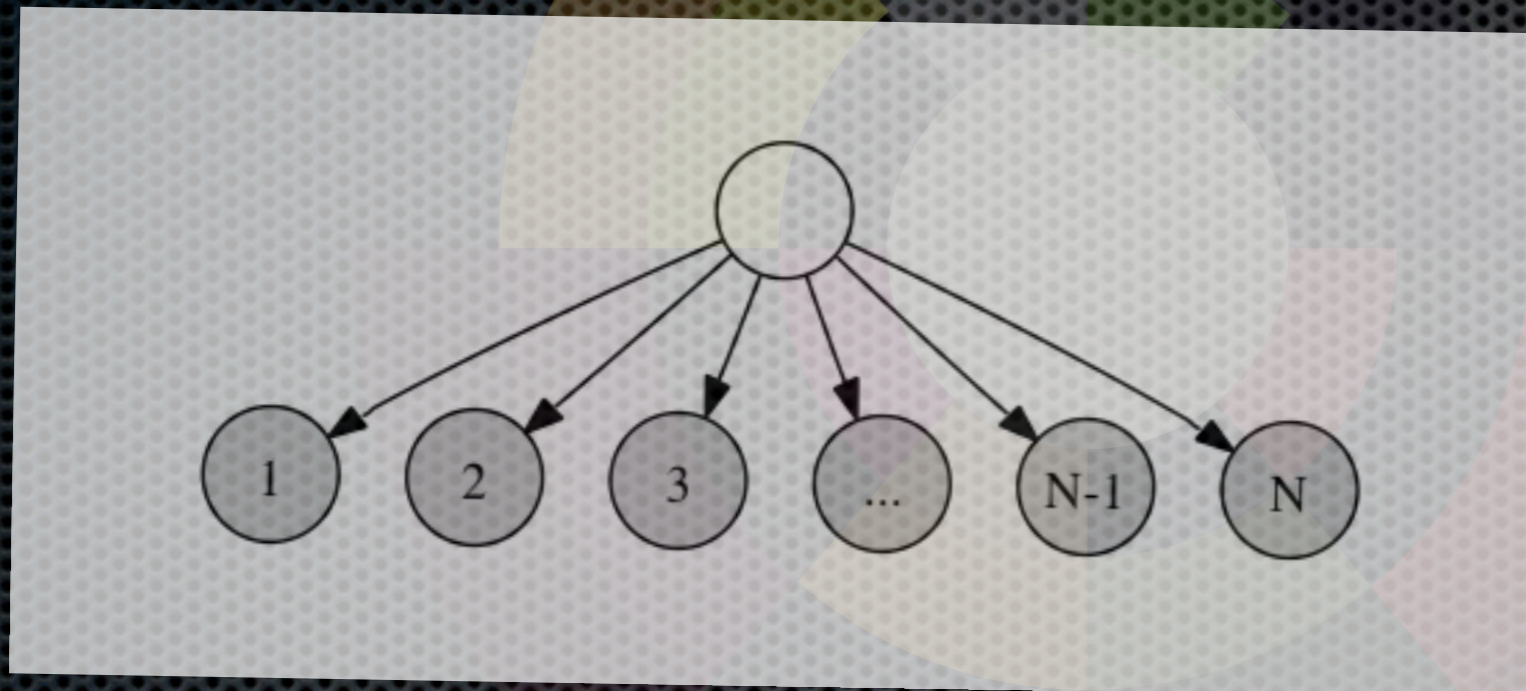


Motivation and Trends

- ✦ Prevalent shift in how people discover information
- ✦ The growth of social media is rampant
- ✦ Social analytics are increasingly important
 - ✦ Online networks can be a powerful force for good
 - ✦ How can we mobilize our networks?

Broadcast Network Utility

- ✦ Sarnoff's law



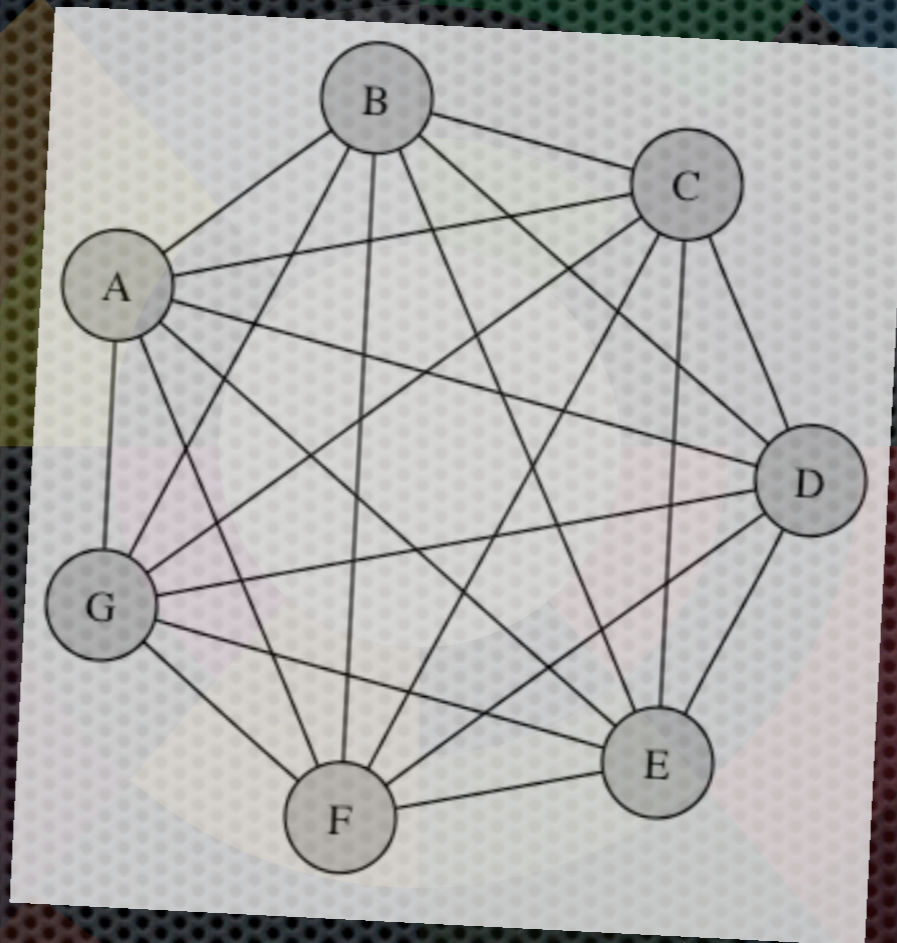
Broadcast Network
utility proportional to the
number of viewers N

- ✦ Example: television or radio network

Interesting Statistic
31.1 million US TV viewers
watched Michael Jackson's memorial
(millions more watched online)

Peer Connecting Network Utility

- ✦ Metcalfe's Law



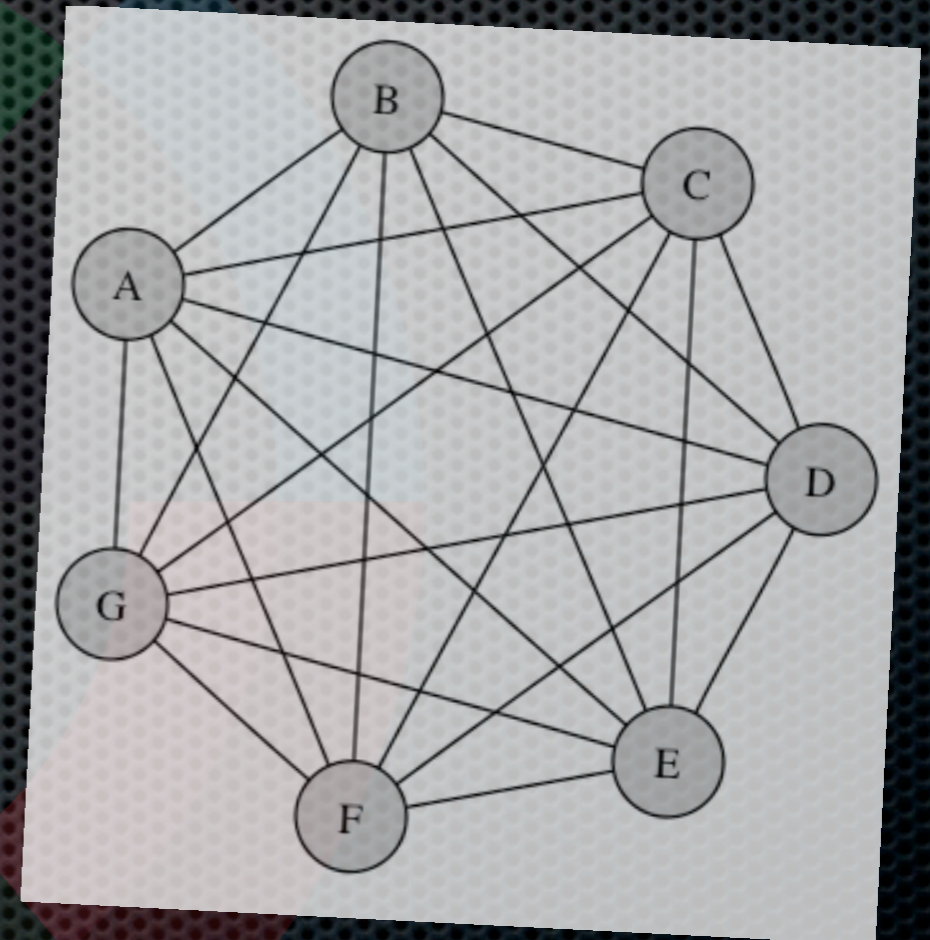
Peer Connecting Network
utility proportional to $N(N-1)/2$

- ✦ Example: telecommunications network

Group Forming Network Utility

- Reed's Law

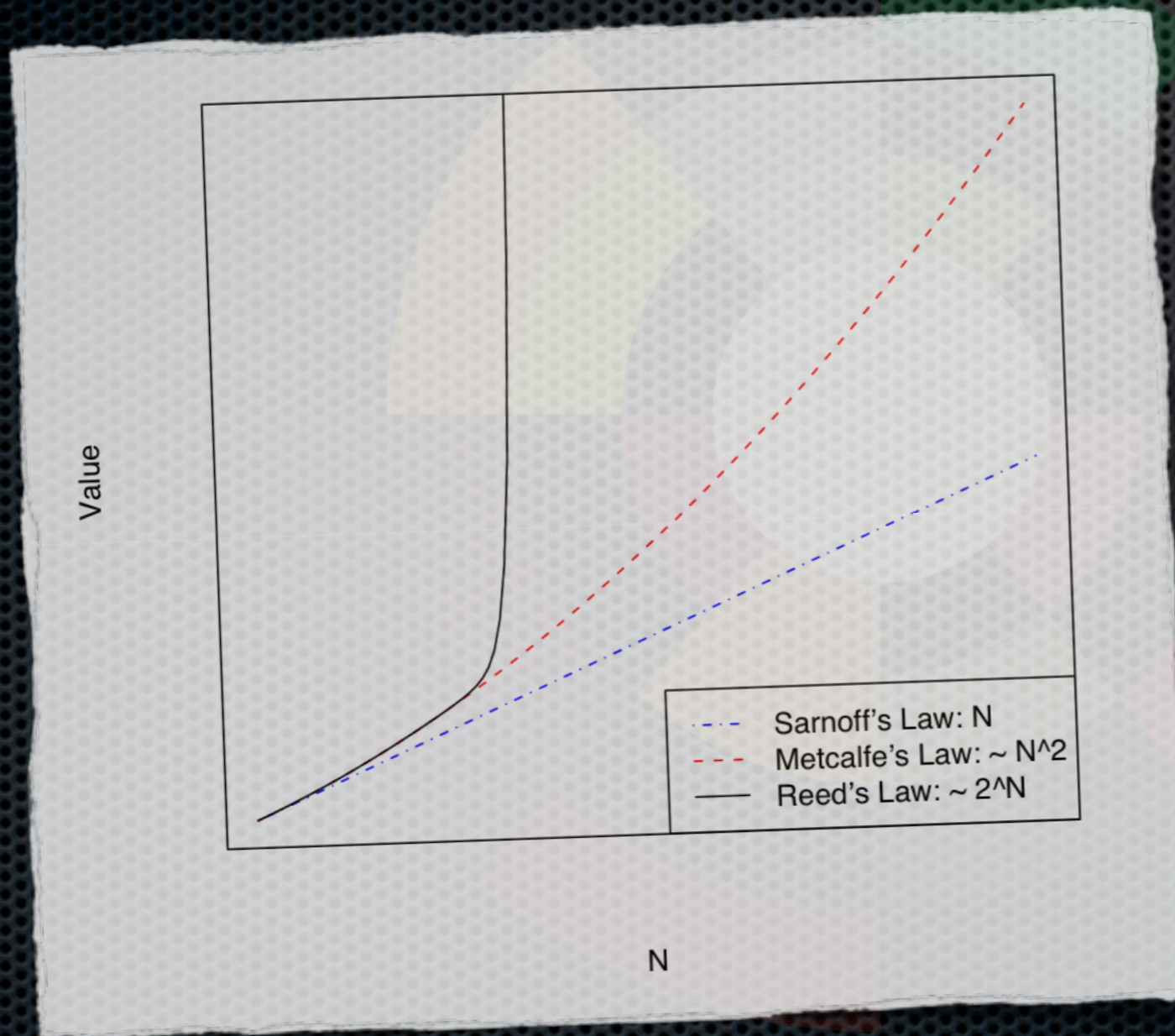
Size	Groups	#
2	{A,B}, {A,C}, {A,D}, ..., {F,G}	21
3	{A,B,C}, {A,B,D}, ..., {E,F,G}	35
4	{A,B,C,D}, {A,B,D,E}, ..., {D,E,F,G}	35
5	{A,B,C,D,E}, ..., {C,D,E,F,G}	21
6	{A,B,C,D,E}, ..., {B,C,D,E,F,G}	7
7	{A,B,C,D,E,F,G} (<i>shown above</i>)	1
Total groups:		120



- Example: online social networks

Group Forming Network (GFN)
utility proportional to 2^N

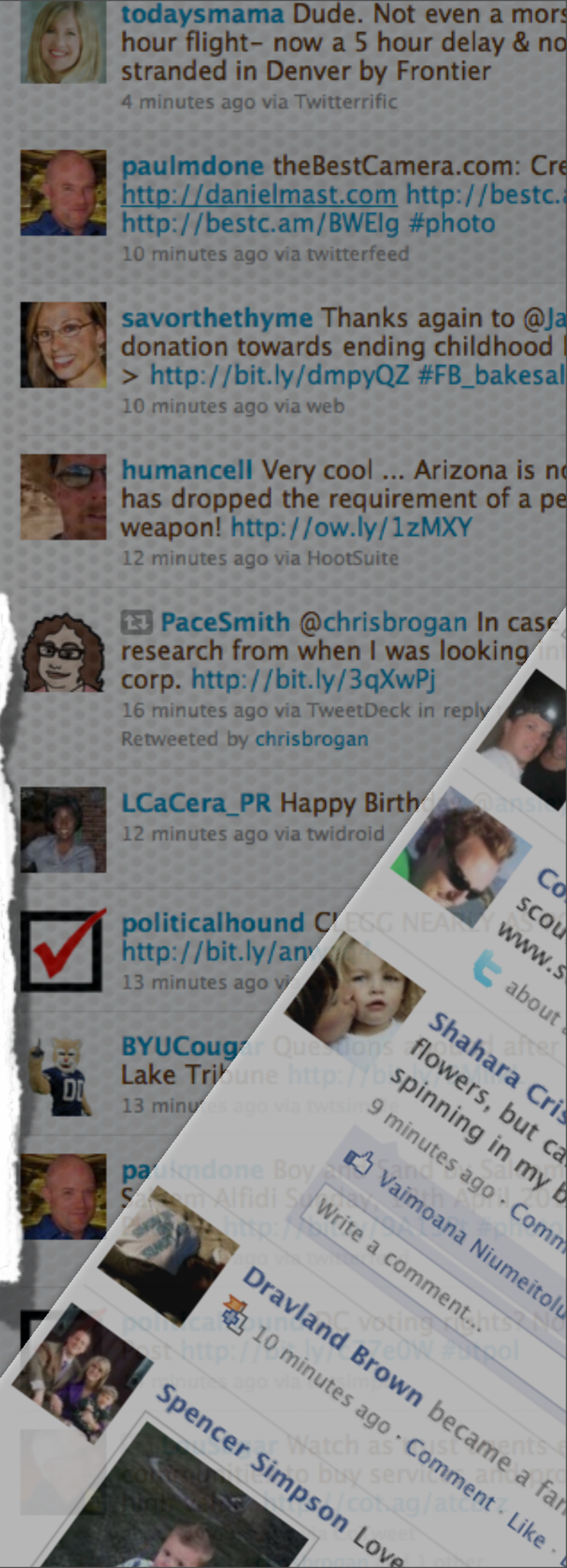
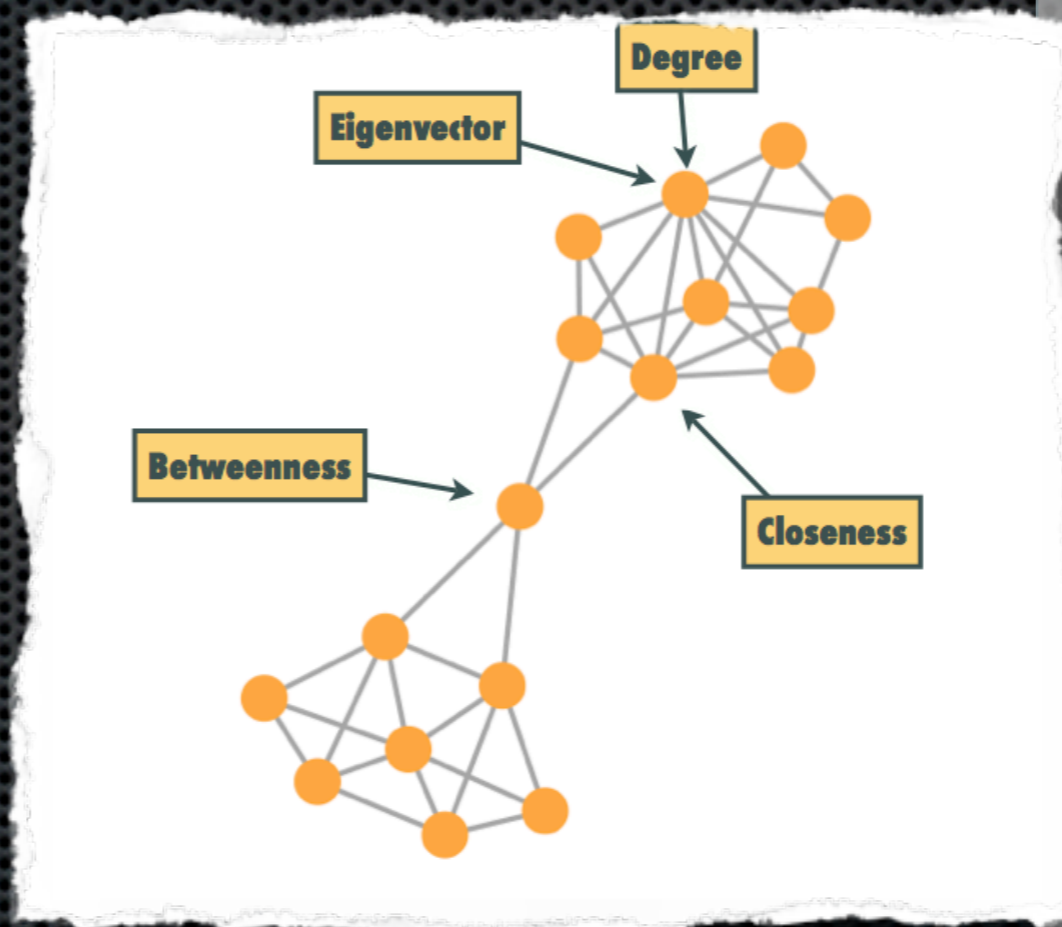
Utility Comparison



Reed's Law
The utility associated with facilitating group affiliations (Reed's Law) far surpasses both the other laws

Social Stream Data

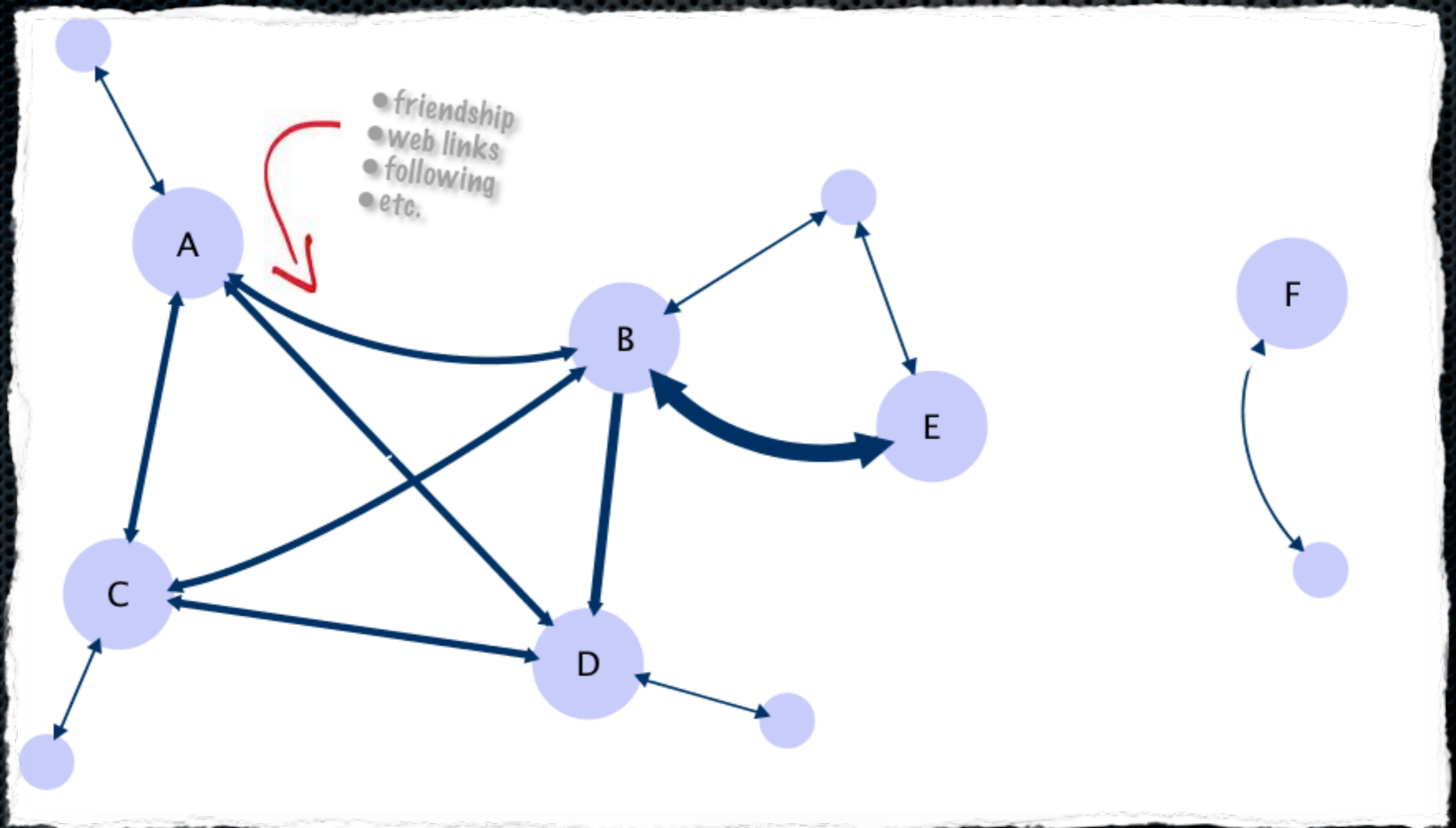
- ✦ Twitter
- ✦ Facebook
- ✦ Blogosphere
- ✦ YouTube
- ✦ Flickr



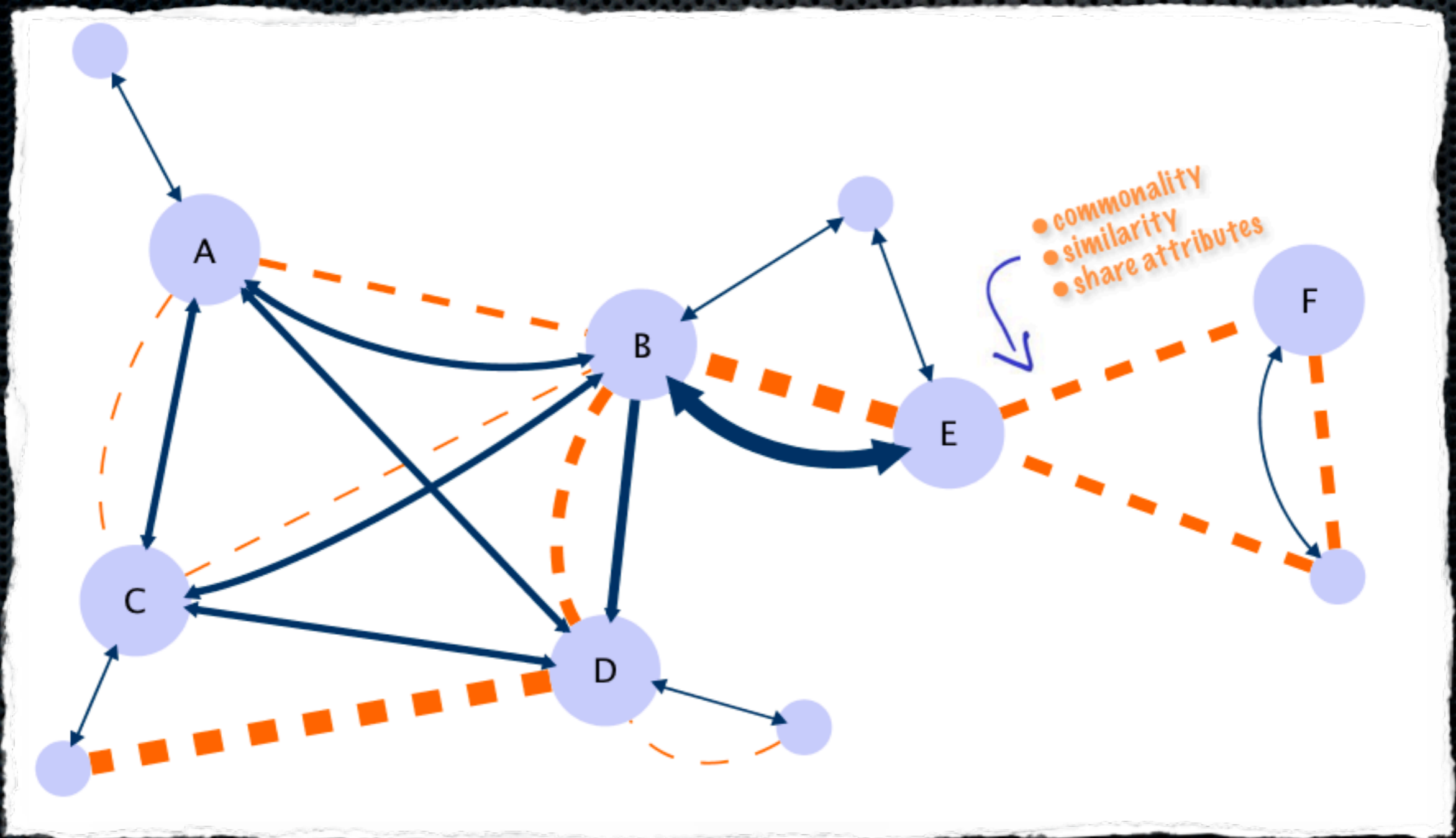
Types of Network Links

- ✦ **Explicit Link** (Relationship)
 - ✦ Direct knowledge, interaction, or communication
 - ✦ Ex. friends, web links, and club members
 - ✦ Explicit Social Networks (ESNs)
- ✦ **Implicit Link** (Affinity)
 - ✦ Inherent similarities or affinities
 - ✦ Ex. attributes, hobbies, interests, and background
 - ✦ Implicit Affinity Networks (IANs)

Social Analytics Framework

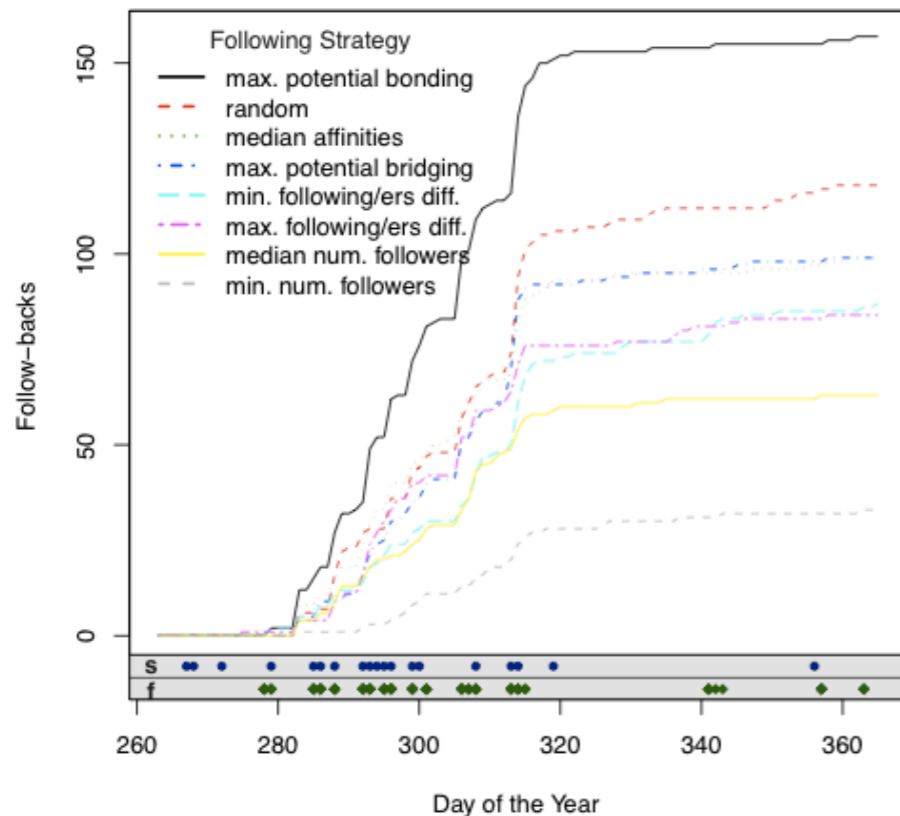


Social Analytics Framework



Experiment - Twitter Networking

rank _f	strategy	following	follow-backs	↓ followers	rejects	churn	follow _{total}	follower _{total}
1	<i>max. potential bonding</i> (A)	500	158 (32%)	202 (40%)	12	127	512	329
2	<i>max. following/ers diff.</i> (F)	500	84 (17%)	172 (34%)	12	324	512	496
3	<i>random</i> (D)	500	118 (24%)	154 (31%)	20	103	520	257
4	<i>median affinities</i> (C)	500	99 (20%)	123 (25%)	25	93	525	216
5	<i>max. potential bridging</i> (B)	500	99 (20%)	120 (24%)	25	91	525	211
6	<i>min. following/ers diff.</i> (E)	500	87 (17%)	99 (20%)	50	55	550	154
7	<i>median num. followers</i> (G)	500	63 (13%)	86 (17%)	31	51	531	137
8	<i>min. num. followers</i> (H)	500	33 (07%)	42 (08%)	79	29	579	71
9	<i>follow nobody</i> (I)	0	0 (—%)	3 (—%)	0	24	0	27



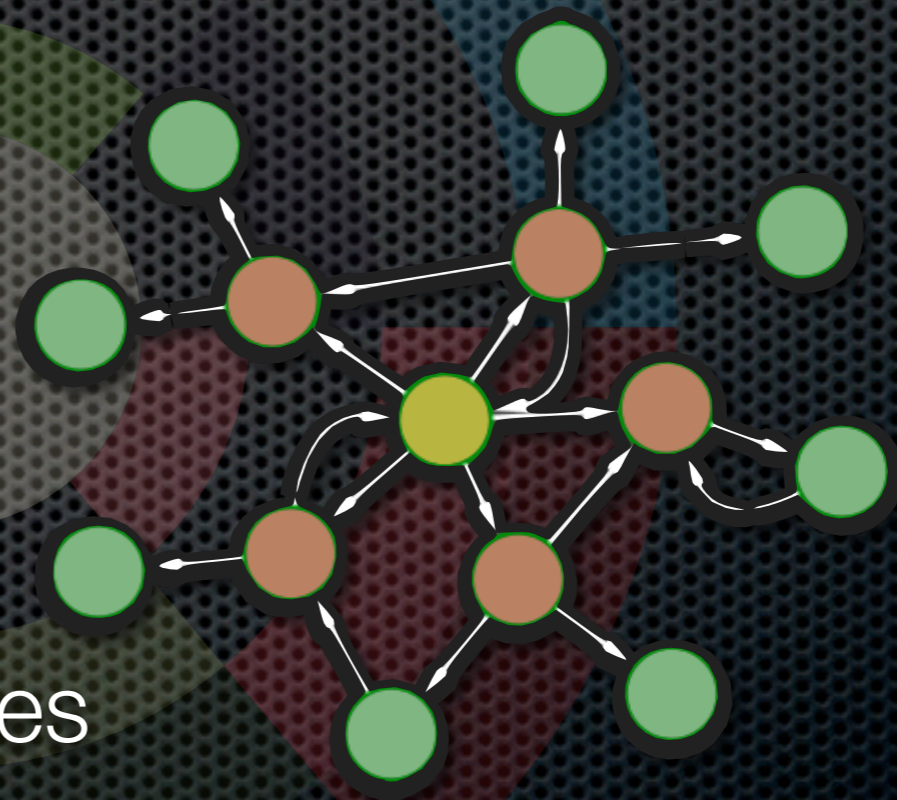
strategy	significantly different
(A) <i>max. potential bonding</i>	B, C, E, F, G, H
(B) <i>max. potential bridging</i>	A, H
(C) <i>median affinities</i>	A, G, H
(D) <i>random</i>	G, H
(E) <i>min. following/ers diff.</i>	A, H
(F) <i>max. following/ers diff.</i>	A, H
(G) <i>median num. followers</i>	A, D, C
(H) <i>min. num. followers</i>	A, B, C, D, E, F

Table 3: **Followback-to-Following:** Pairwise Proportion Test Results. ($\alpha = 0.01$, Bonferroni corrected p -values)



Experiment - Blogosphere

- ✦ Niche focused
- ✦ Bonding beneficial
- ✦ Track overtime
- ✦ Identify uncaptialized affinities



Conclusions

- ✦ The growth of social media is rampant
- ✦ Technology enabling Group Forming Networks (GFNs)
- ✦ Social capital framework provides useful social analytics
- ✦ We can test sociological hypotheses

Questions & Comments

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