

A Computational Framework for Social Capital in Online Communities

Matthew Smith

<http://m.smithworx.com>

Department of Computer Science

<http://dml.cs.byu.edu>

Brigham Young University



DML

Data Mining Lab

PhD Dissertation Defense, April 2011

Introduction

- 📌 Online communities increasingly important
- 📌 Prevalent shift in how people discover information
- 📌 Large social networks are dynamic and complex
- 📌 Rich social network data is becoming available
- 📌 Social capital within these networks is poorly understood
- 📌 **Computational Social Science**





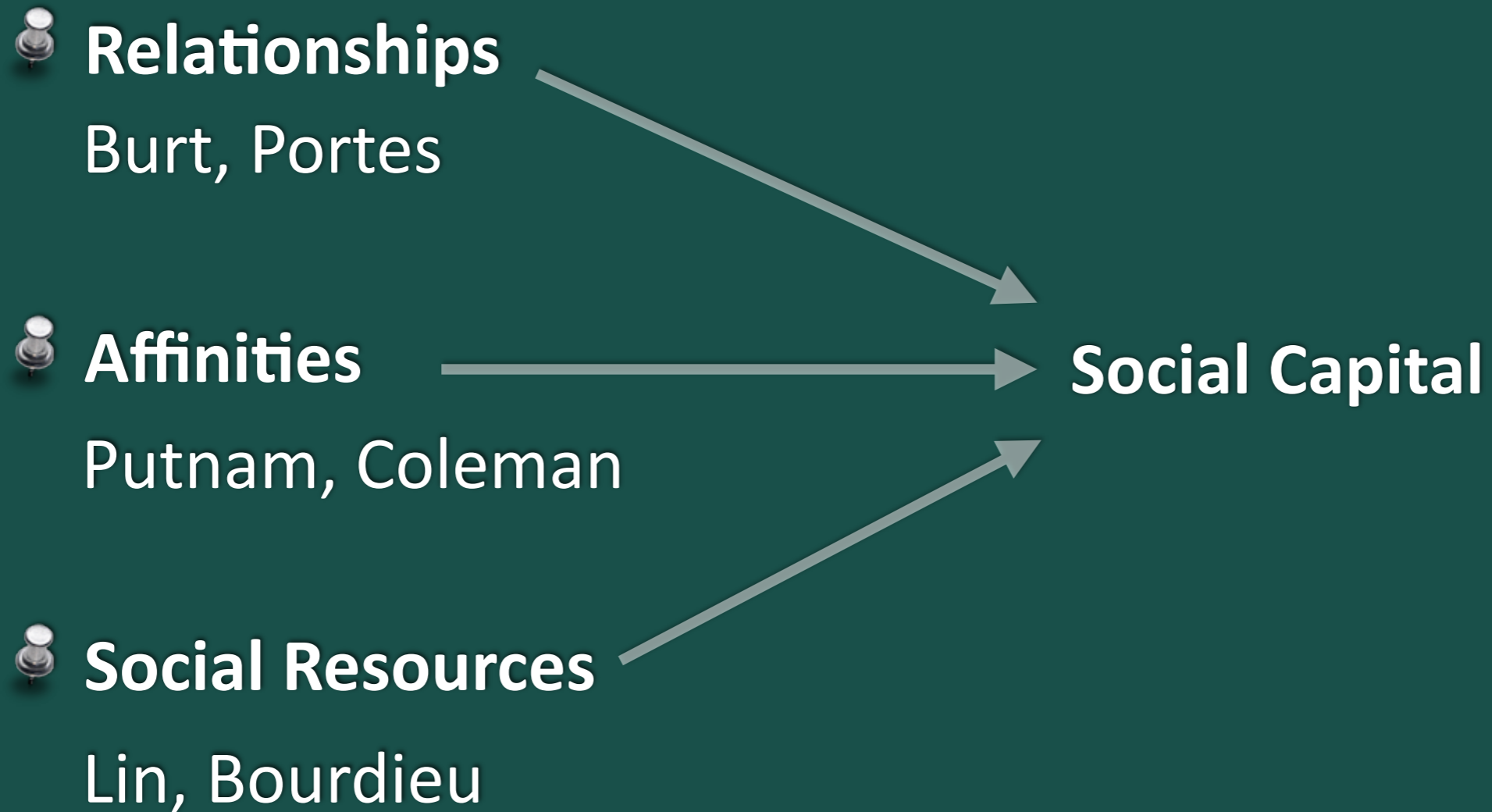
Social Capital Framework

Social Capital

- 📌 **The value of social networks.** More specifically,
 - 📌 ***Bonding*** similar people together (Putnam)
 - 📌 ***Bridging*** diverse people together (Putnam)
 - 📌 **Access to and *use of resources*** embedded in social networks (Lin)





Social Capital





Relationships



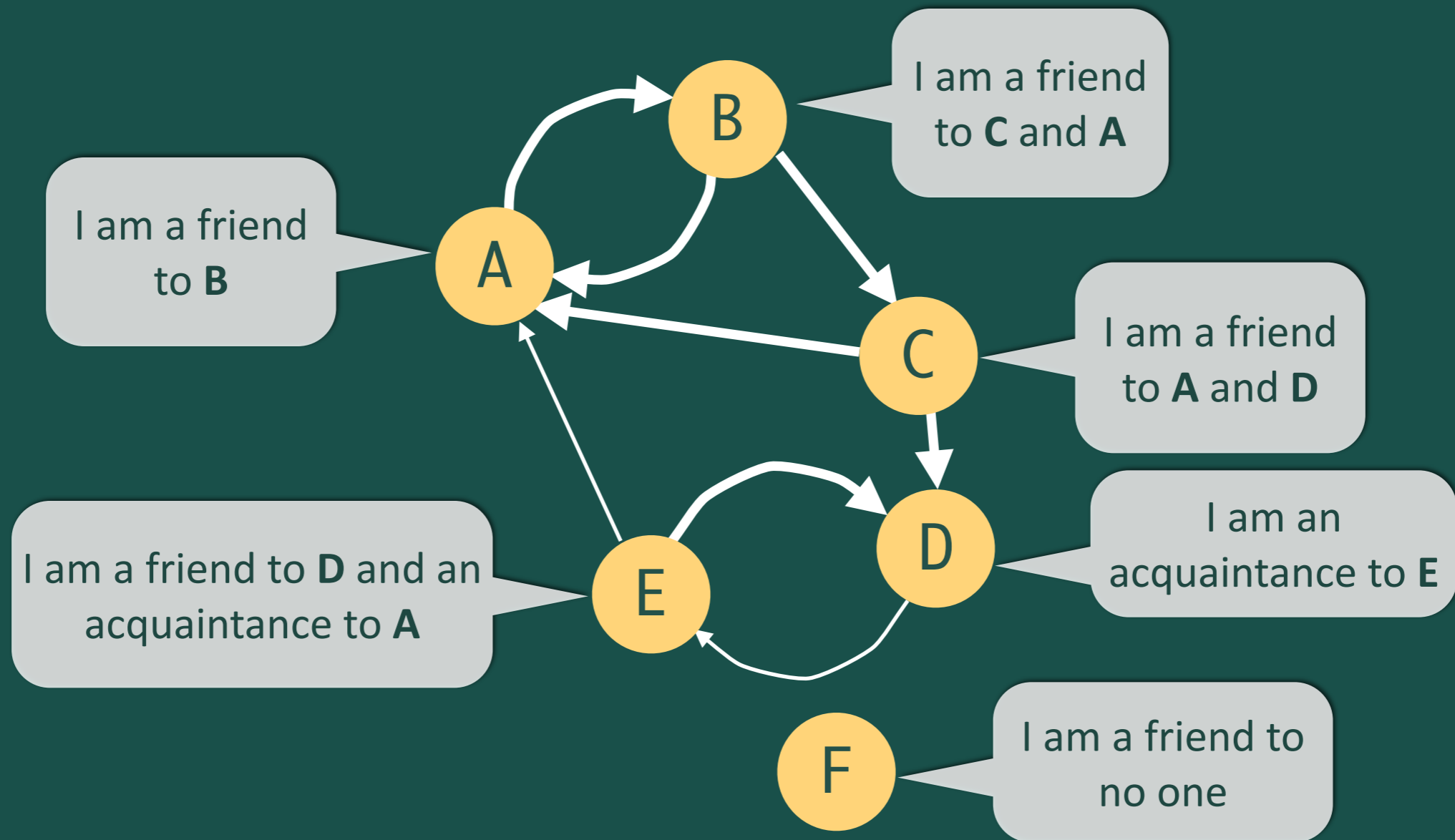
-  Direct knowledge, interaction, or communication
-  Ex. behavioral interaction, evaluation of one by another (e.g., friends, web links), formal (e.g., co-worker), biological (e.g., sibling)



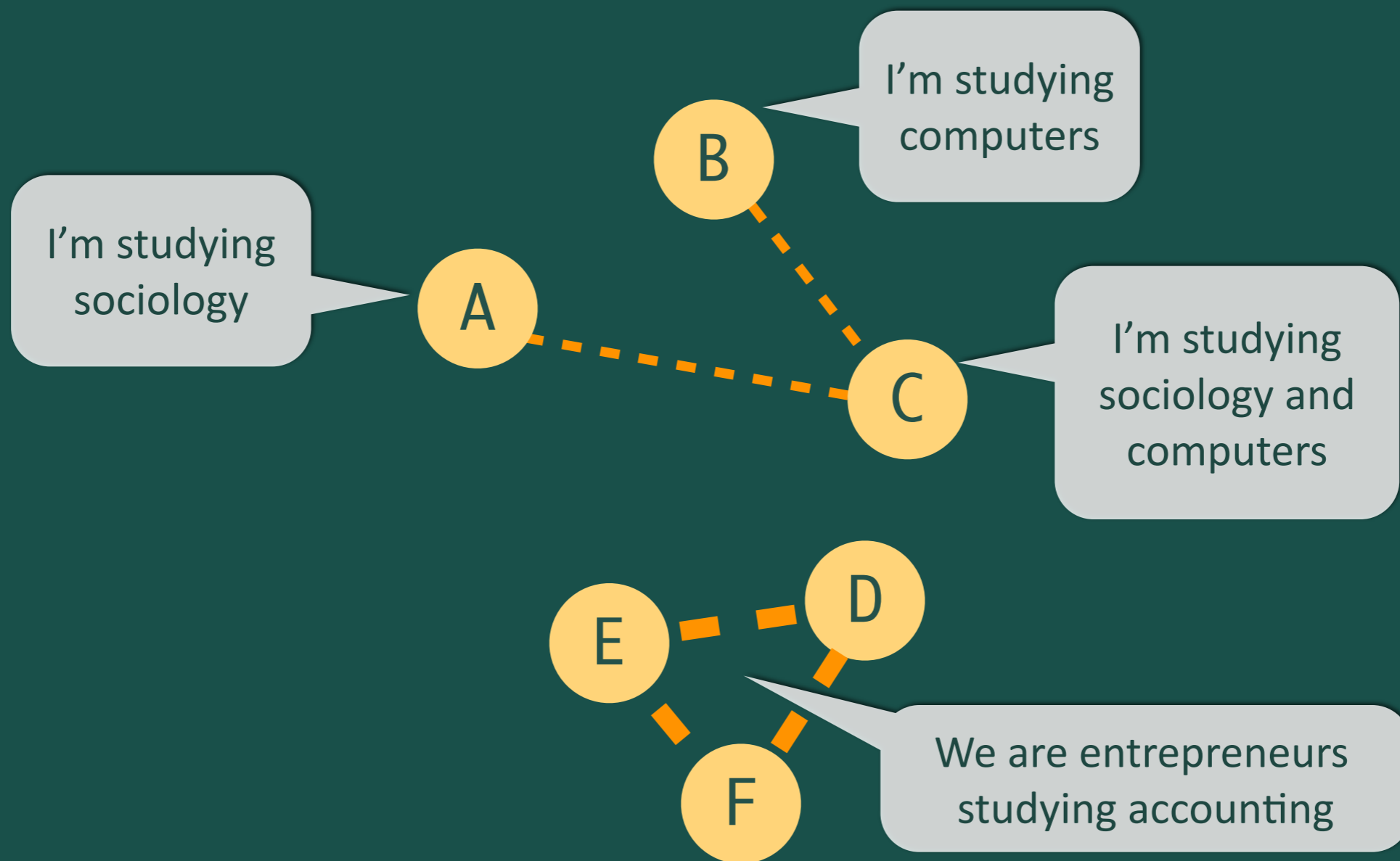
-  Inherent similarities or affinities
-  Ex. attributes, hobbies, interests, background, etc.



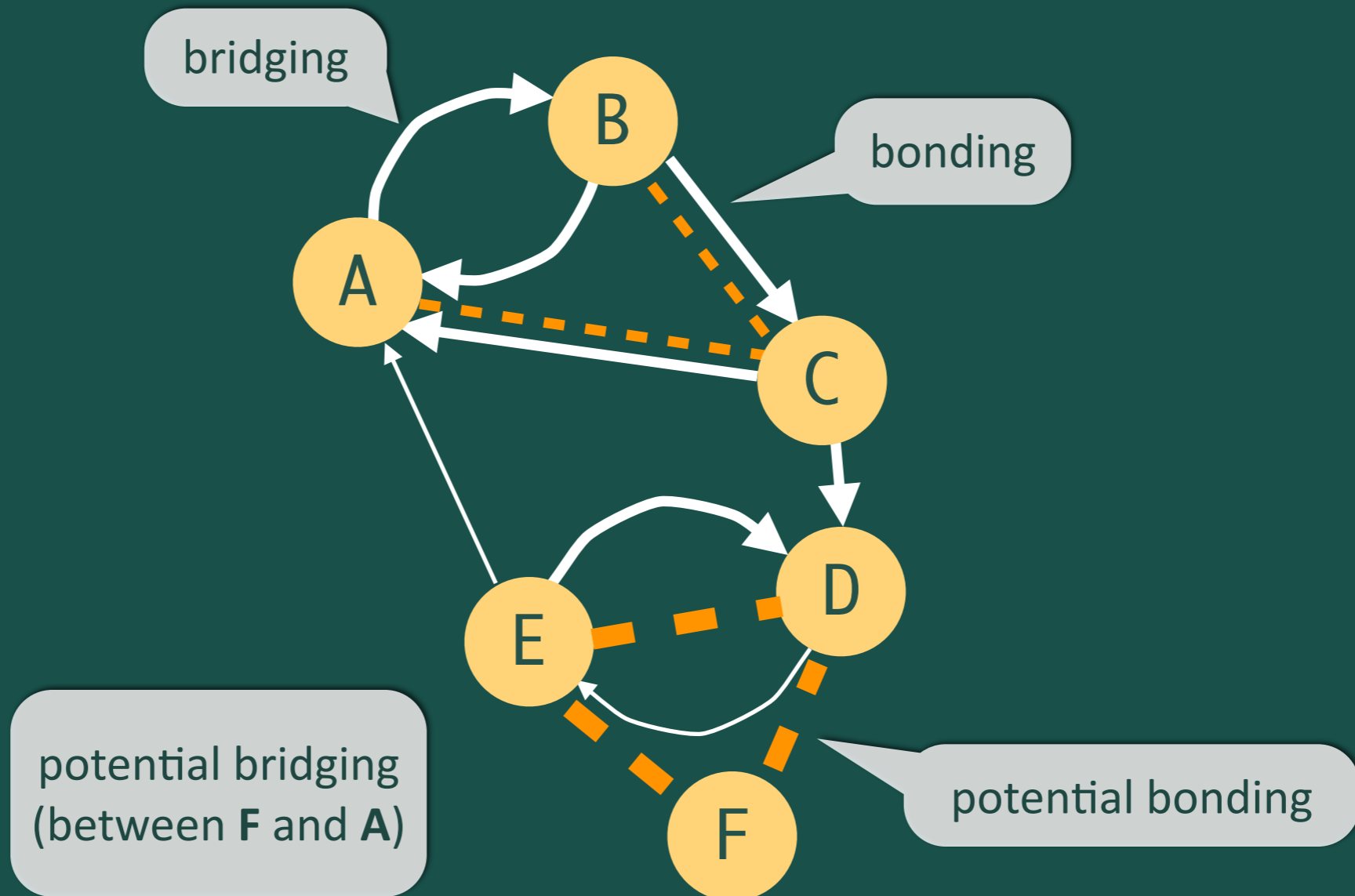
Explicit Social Network (ESN)



Implicit Affinity Network (IAN)



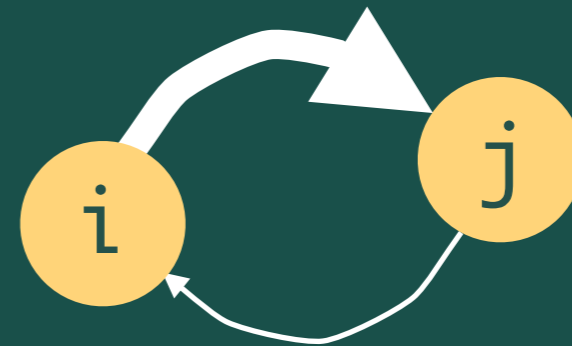
Hybrid Network



Relationships

📌 Directed

📌 Strength



📌 *Initialization*: Behavioral interaction, evaluation of one person by another, formal relationships, biological relationships; base on affinities

📌 *Dynamic*: interactions, time decay



Affinities

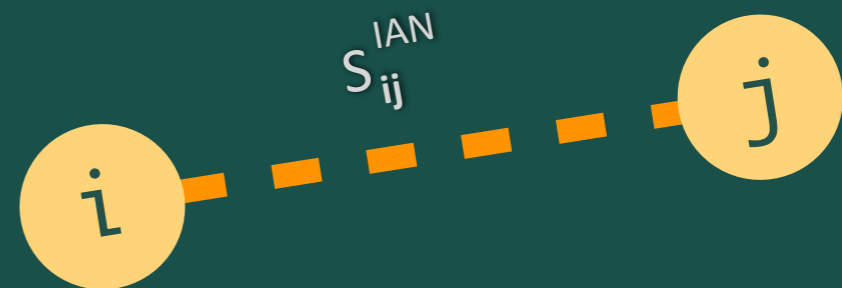
- 📌 The set of attributes exposed by individual j

$$A_j$$

- 📌 The affinity strength between individuals i and j

$$s_{ij}^{IAN} = \frac{A_i \cap A_j}{A_i \cup A_j}$$

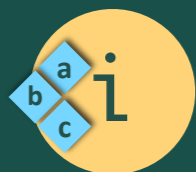
Jaccard Index
(set similarity)



Resources



- 📌 ***Social resource***: a specific asset, material or symbolic, available through social connections within a network
- 📌 ***Characteristics***: (e.g., exhaustible, returnable, quantifiable, durable)
- 📌 ***Value***: assigned by the individual (often dependent on characteristics)
- 📌 ***Possessed or Sought***



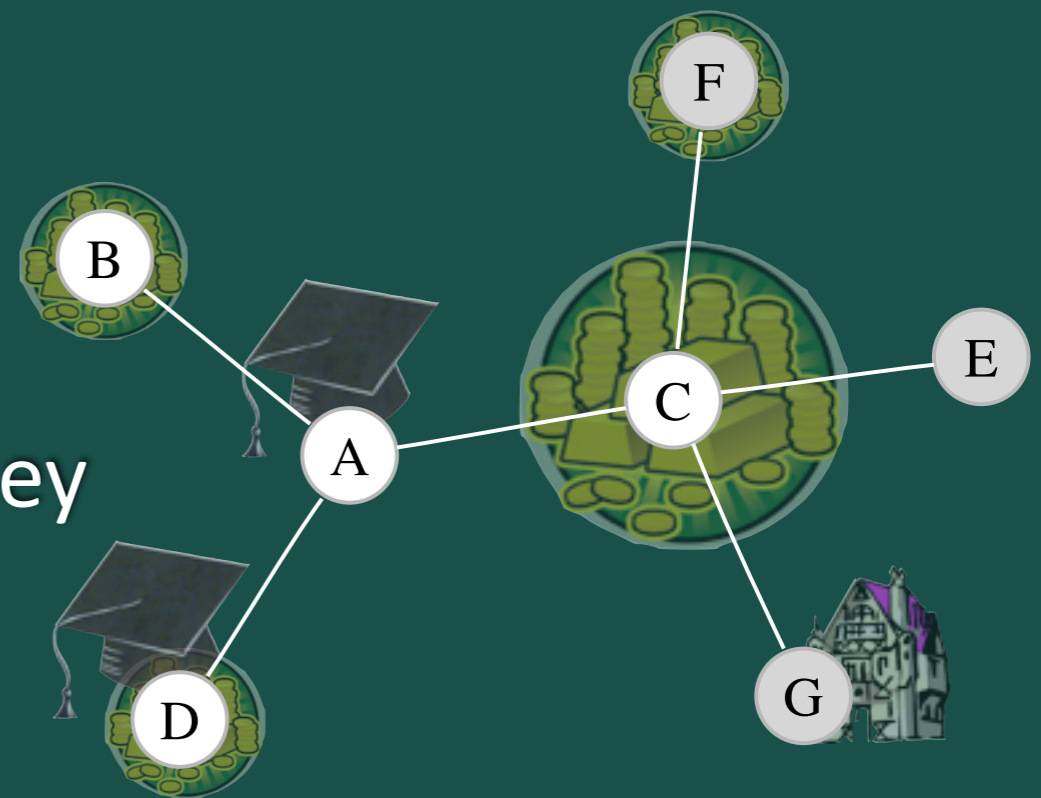
Social Resource Examples

Material Goods

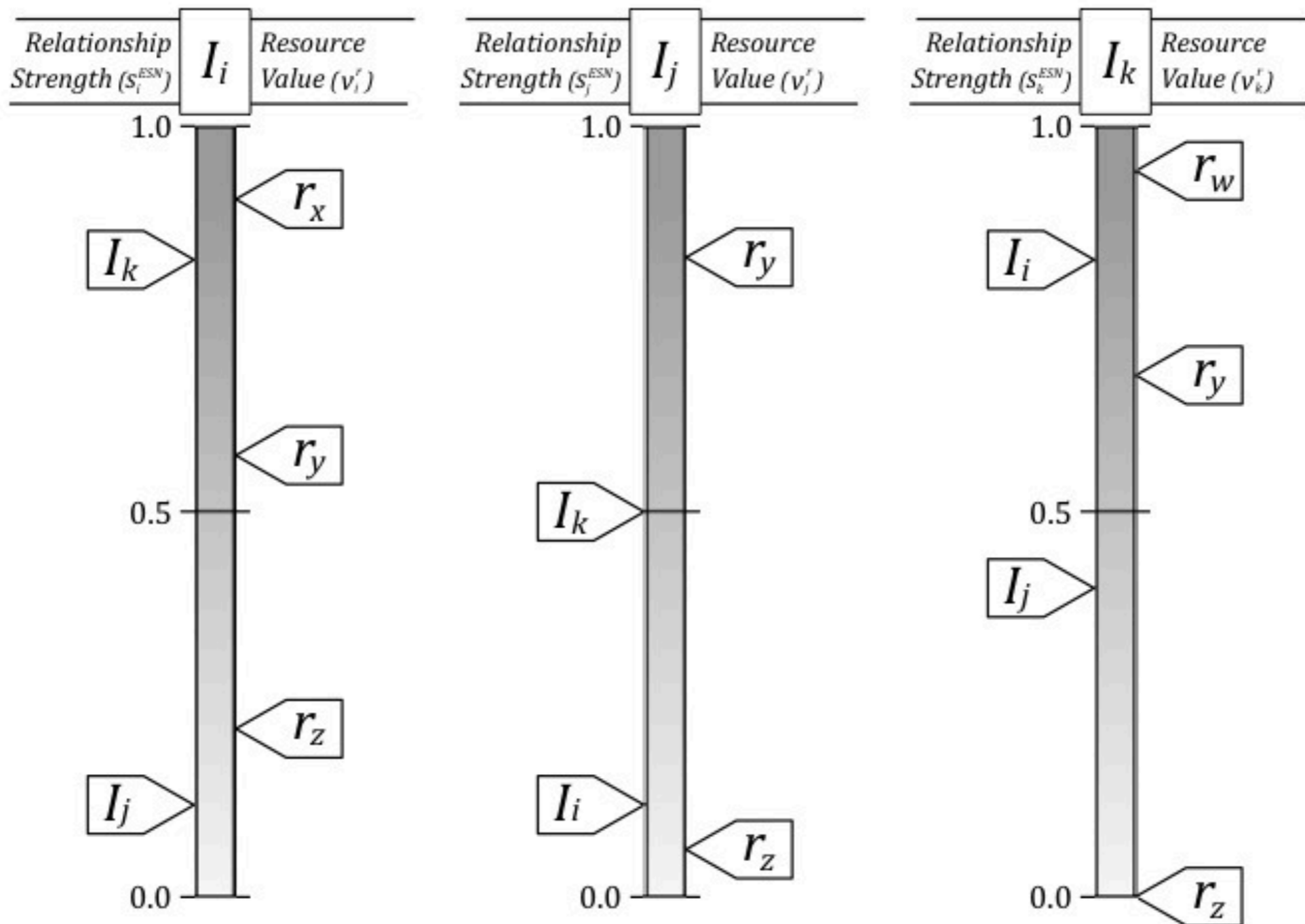
- land, houses, car, and money

Symbolic Goods

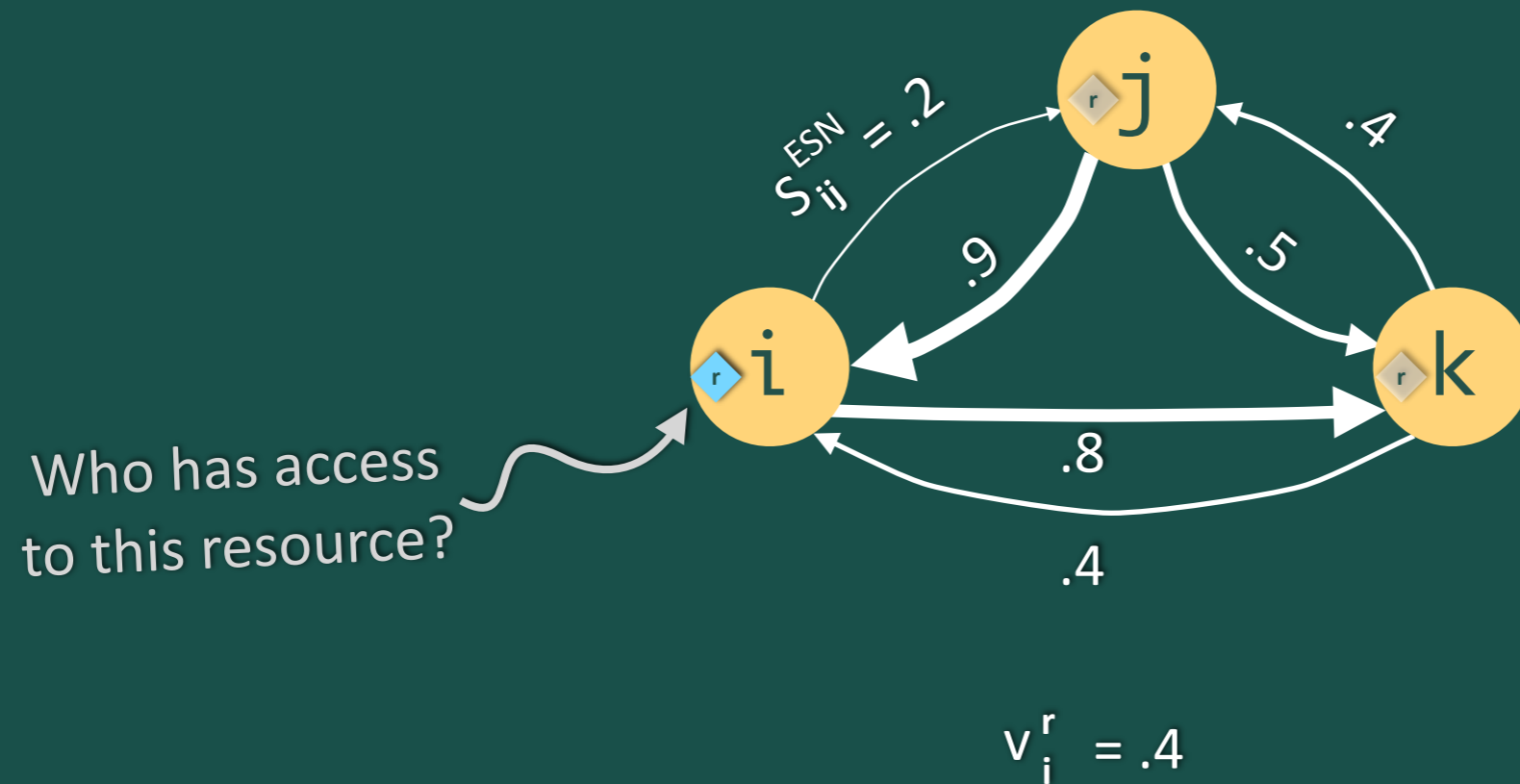
- education, memberships in clubs, honorific degrees, nobility or organizational titles, family name, reputation, or fame



Access to Resources



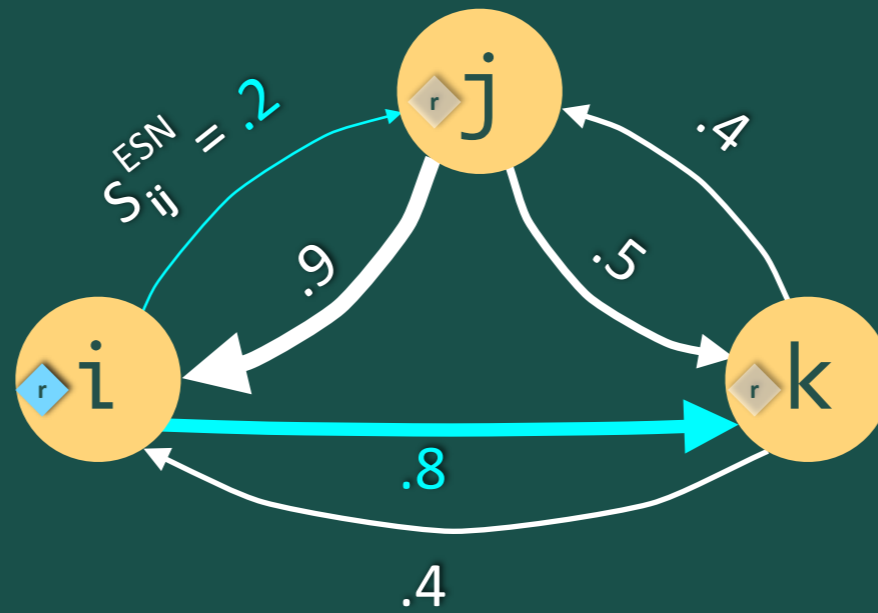
Access to Resources



$$access_j(i, r) = \begin{cases} True & \text{if } s_{ij}^{ESN} \geq v_i^r \\ False & \text{otherwise} \end{cases}$$

Access to Resources

Only individual k
has access



$$v_i^r = .4$$

$$access_j(i, r) = \begin{cases} True & \text{if } s_{ij}^{ESN} \geq v_i^r \\ False & \text{otherwise} \end{cases}$$

Social Capital Computation

- Resources available to individual j from individual i

$$\#res_j(i) = | \{r \in R_i^p : access_j(i, r) = True \wedge C_j(r) = Yes\} |$$

Context Vector

- Individual Social capital

$$sc(j) = \sum_i \#res_j(i)$$

Bonding and Bridging Social Capital

• Bonding and Bridging

$$b(j) = \sum_i \#res_j(i) s_{ij}^{IAN}$$

$$br(j) = \sum_i \#res_j(i) (1 - s_{ij}^{IAN})$$

• Individual

$$sc(j) = b(j) + br(j)$$

• Network

$$sc = \sum_j sc(j)$$

Bonding and Bridging Social Capital

Bonding and Bridging

$$b(j) = \sum_i s_{ij}^{ESN} s_{ij}^{IAN}$$

$$br(j) = \sum_i s_{ij}^{ESN} (1 - s_{ij}^{IAN})$$

Note: used as surrogate when resources are not available

Individual

$$sc(j) = b(j) + br(j)$$

“It is a very interesting implementation of the social capital ideas, especially *bridging and bonding*. I have long thought that the social capital research field would be immensely aided if some of it could be formalized in terms of network theory, and this is **the best job of that that I've seen so far.**”

- Robert Putnam (personal correspondence)

Interactions

- 📌 **Interaction:** a purposeful exchange between individuals
 - 📌 How is it perceived? (+, -, or neutral)
 - 📌 Is a resource involved?

- 📌 Who to interact with?



Interactions

📌 Generalized form

$$i = sel_j(R_i^p, R_j^s, s_{ij}^{ESN}, s_{ij}^{IAN})$$

Resources possessed by i

Resources sought by j

Seek resources in a priority order

📌 Instantiated examples (determined by j 's goals)

$$i = argmin_k(\min_{r \in R_j^s \cap R_k^p} (v_k^r - s_{kj}^{ESN}))$$

$$i = argmin_k(v_k^{rh} - s_{kj}^{ESN})$$

Get sought after resources as quickly as possible

$$i = argmax_k(|R_k^p| + s_{kj}^{IAN})$$

Seek resources from who has many and is most similar to you

Dynamic Interaction Function

$$\Delta s_{ji}^{ESN}(\text{event})$$

Event	Δs_{ji}^{ESN}	Δs_{ij}^{ESN}
j gives resource r to i	−	+
j receives resource r from i	+	−
j engages in a (resourceless) interaction with i	−	+
j receives a (resourceless) interaction from i	+	−

Implementation

NetLogo

The image displays the NetLogo interface for a network simulation. The central area shows a network of 10 nodes (represented by human figures) connected by 90 relationships (green lines). The nodes are arranged in a circular pattern, and the relationships form a dense web.

Control Panel (Left):

- network-choice: fully-connected
- initial-resource-distribution: disparately
- number-of-nodes: 10
- number-of-resources: 10
- number-of-unique-resour...: 26
- global-hold-onto...: 5
- label-nodes-by: resources
- scale-nodes-by: social capital
- network-layout: circle
- drag?: Off
- show-resource-flow?: Off
- show-edge-labels?: Off
- global-strength-thre...: 0.50
- edge-strength-thres...: 0.50
- redraw: R, step: P, go: G
- go-by-round?: Off
- shuffle-turn-o...: Off
- take-resources?: Off
- give-resources?: Off
- turn-interacti...: Off
- verbose?: Off

Monitoring Panel (Right):

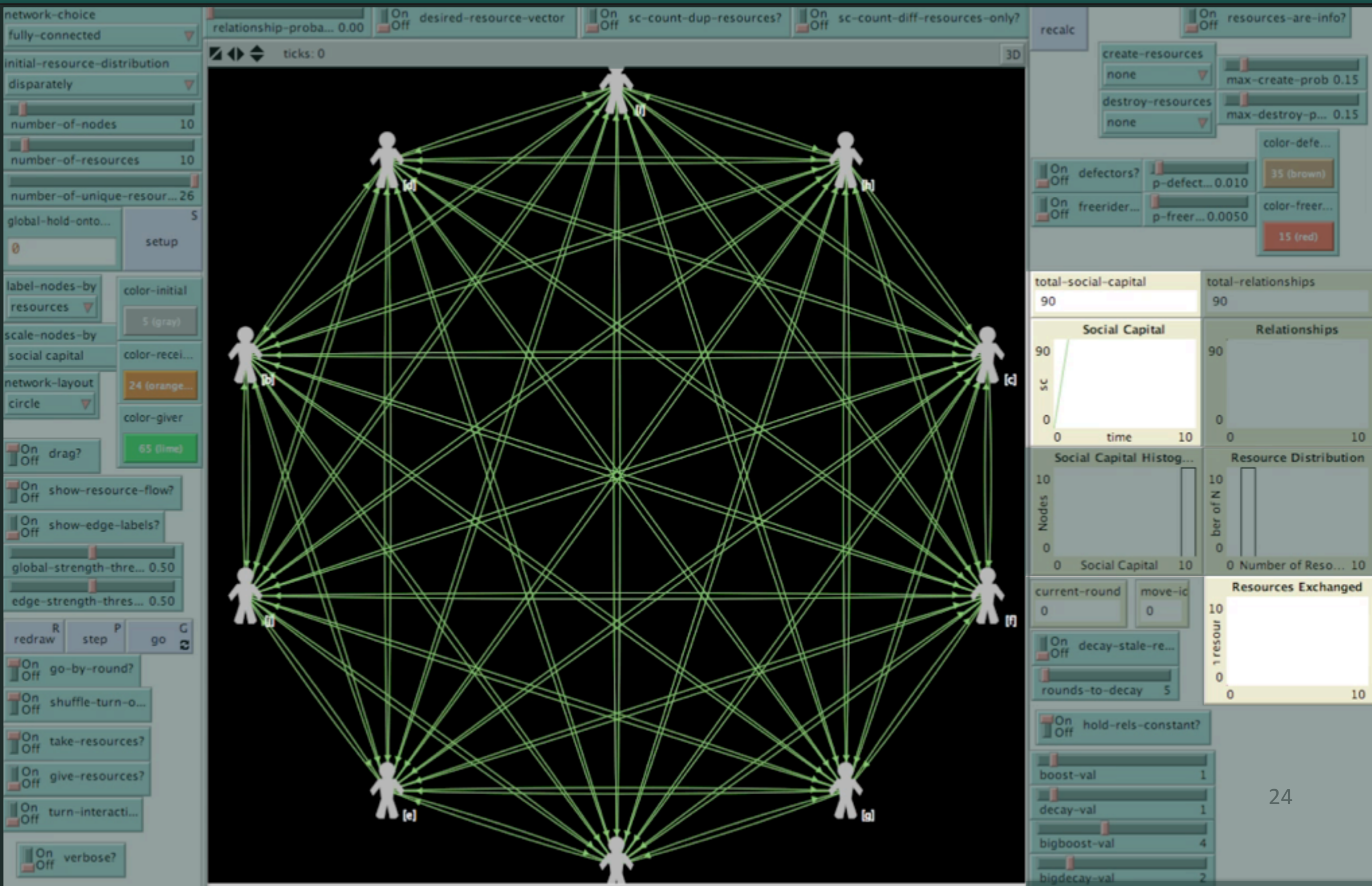
- total-social-capital: 90
- total-relationships: 90
- Social Capital Histogram: Nodes vs Social Capital
- Resource Distribution: ber of N vs Number of Reso...
- Resources Exchanged: 1 resour vs 10
- current-round: 0, move-id: 0
- decay-stale-re...: Off
- rounds-to-decay: 5
- hold-rels-constant?: Off
- boost-val: 1
- decay-val: 1
- bigboost-val: 4
- bigdecay-val: 2

Top Panel:

- relationship-proba...: 0.00
- desired-resource-vector: Off
- sc-count-dup-resources?: Off
- sc-count-diff-resources-only?: Off
- resources-are-info?: Off
- create-resources: none, max-create-prob: 0.15
- destroy-resources: none, max-destroy-p...: 0.15
- defectors?: Off, p-defect...: 0.010
- freerider...: Off, p-freer...: 0.0050
- color-defe...: 35 (brown)
- color-freer...: 15 (red)

Simulation

Community & defectors



Publications (Framework)

- 📌 **Implicit Affinity Networks**
(published in *Proceedings of the 17th Annual Workshop on Information Technologies and Systems*)
- 📌 **Implicit Affinity Networks and Social Capital**
(published in *Information Technology and Management - Journal*)
- 📌 **Measuring and Reasoning about Social Capital**
(submitted to *Social Networks*, 2011)







Case Studies

Blogosphere



Blogosphere Experiment

Focus

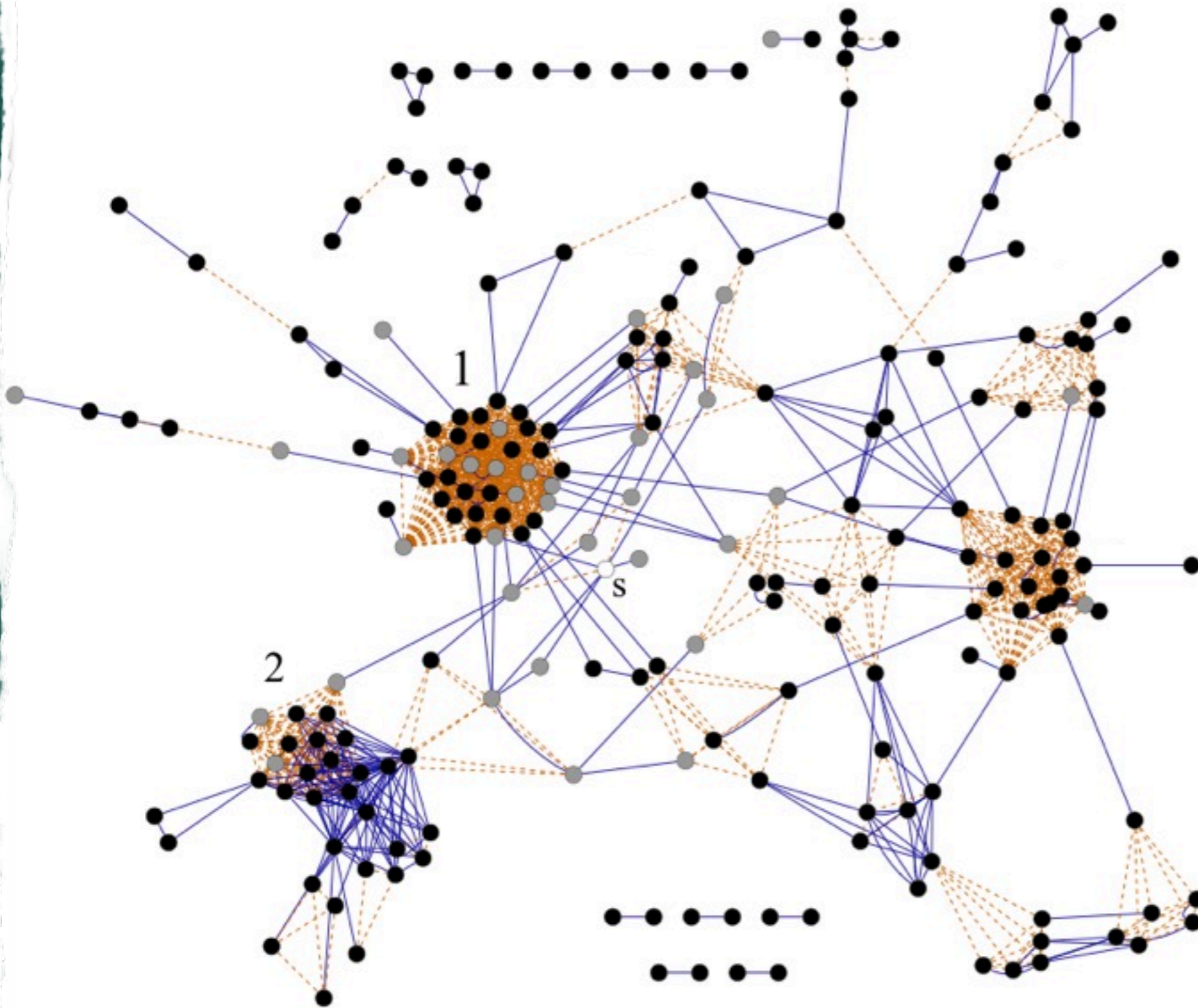
-  Blogger communities centered around topics
-  Started with Scoble's blog

Data

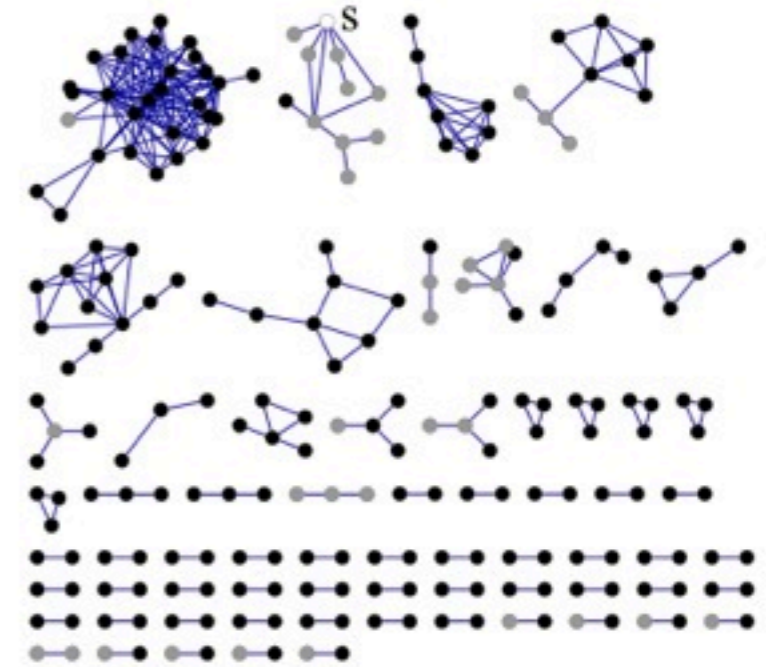
-  13 million blog entires, 38K+ blogs
-  July 2006 - July 2007 (1 year)



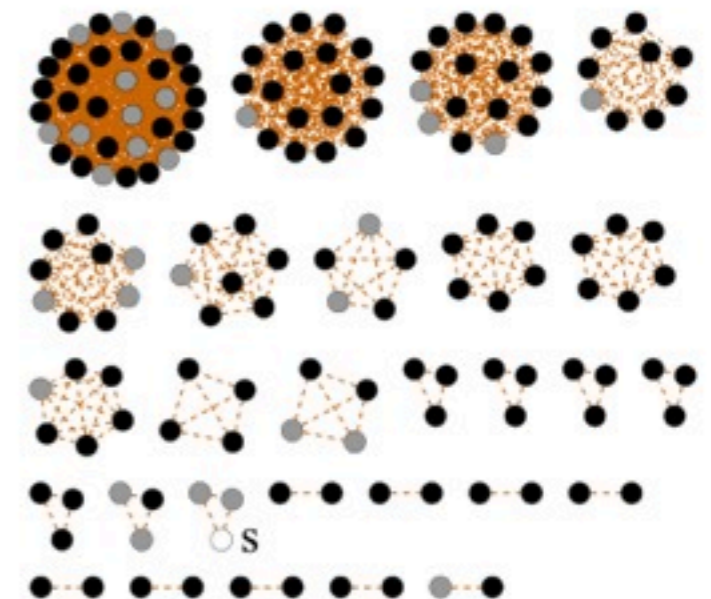
Hybrid Network - Blogosphere



ESN



IAN



Twitter



Twitter Experiment

- 📌 Leverages the Framework to empirically test:
 - 📌 “Bonding is more likely to occur than bridging” - Lin 01
 - 📌 “Closure is the most obvious force” - Burt 05
 - 📌 Homophily principle: “Birds of a feather flock together”
Smith-Lovin 87 “Similarity begets friendship”
- 📌 *Following users with whom the most affinities are shared (i.e., attempting to bond) produces more follow-backs (i.e., bonding) than other following strategies.*

Results

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

strategy	significantly different
(A) <i>bonding</i>	B, C, E, F, G, H
(B) <i>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 III
 FOLLOWBACKS-TO-FOLLOWING: PAIRWISE PROPORTION TEST
 RESULTS. ($\alpha = 0.01$, BONFERRONI CORRECTED p -VALUES)

Follow-backs Retained

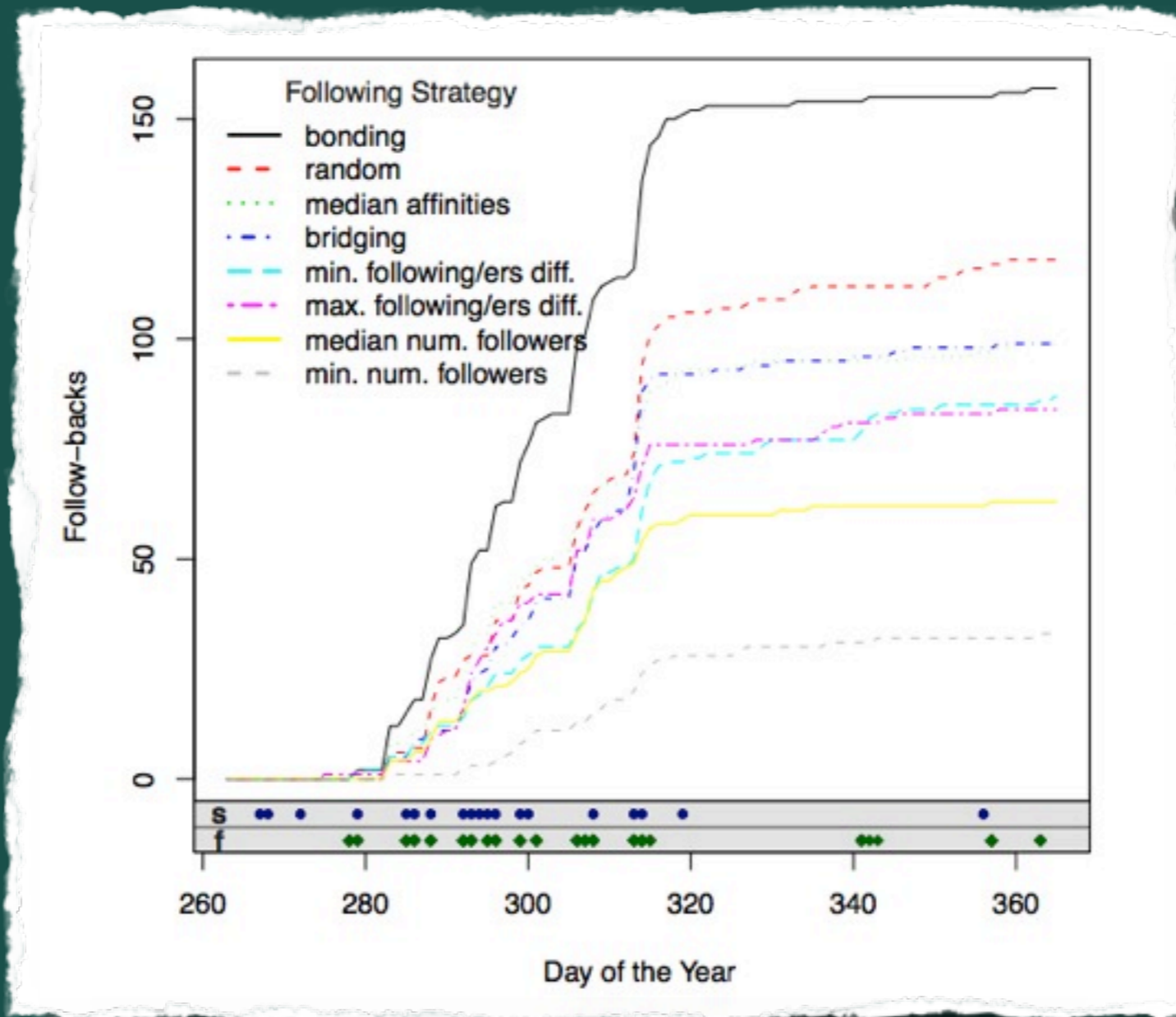


Fig. 2. **Follow-backs Over Time:** Follow-backs obtained by accounts in \mathcal{A} throughout the duration of the study. Days in which following rounds occurred (i.e., accounts in \mathcal{A} followed users in \mathcal{U}) are marked in the row labeled **f**. Days that new status updates were posted to the accounts in \mathcal{A} are marked in the row labeled **s**.

Language Acquisition



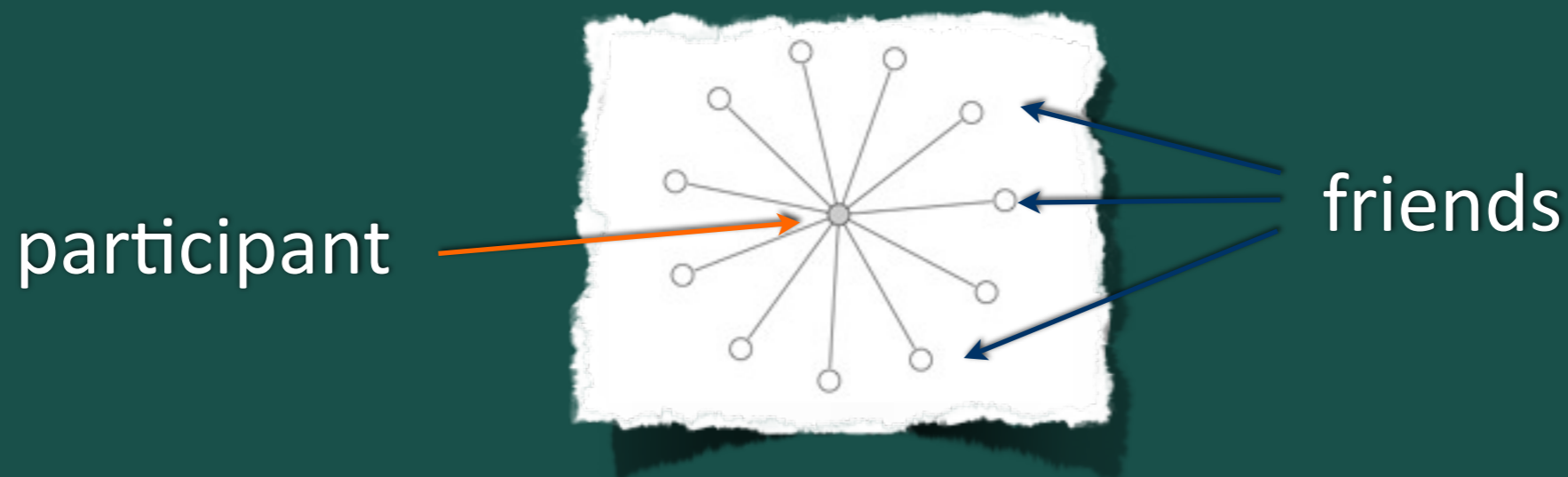
Experimental Setup

- 📌 Study abroad in Japan
- 📌 204 participants
- 📌 Spent an average of 8.4 months in Japan, taking 13.2 hours per week of Japanese language courses in 38 language programs across 22 different cities
- 📌 Language improvement measured (pre-test and post-test)
- 📌 Offline community



Experiment Details

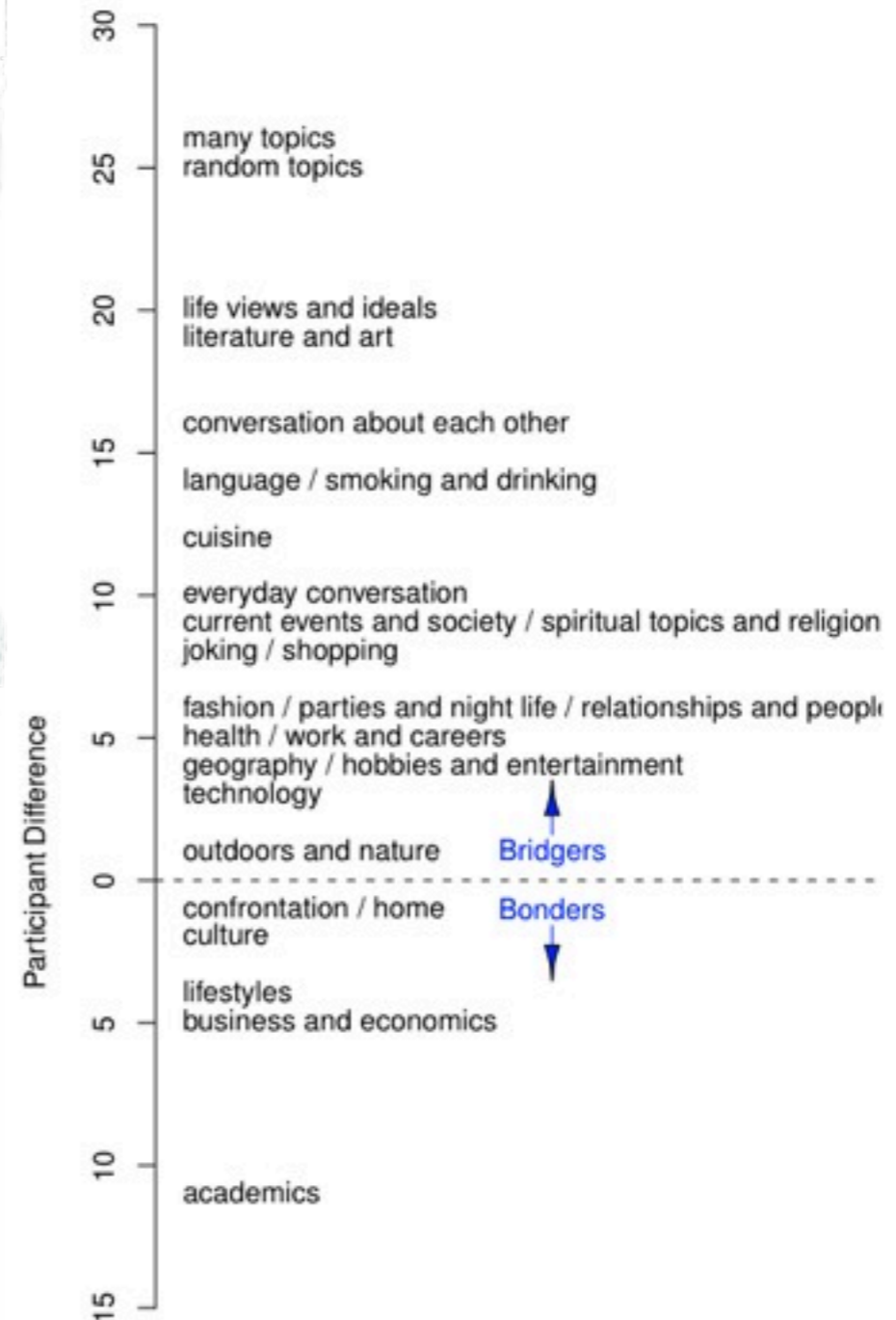
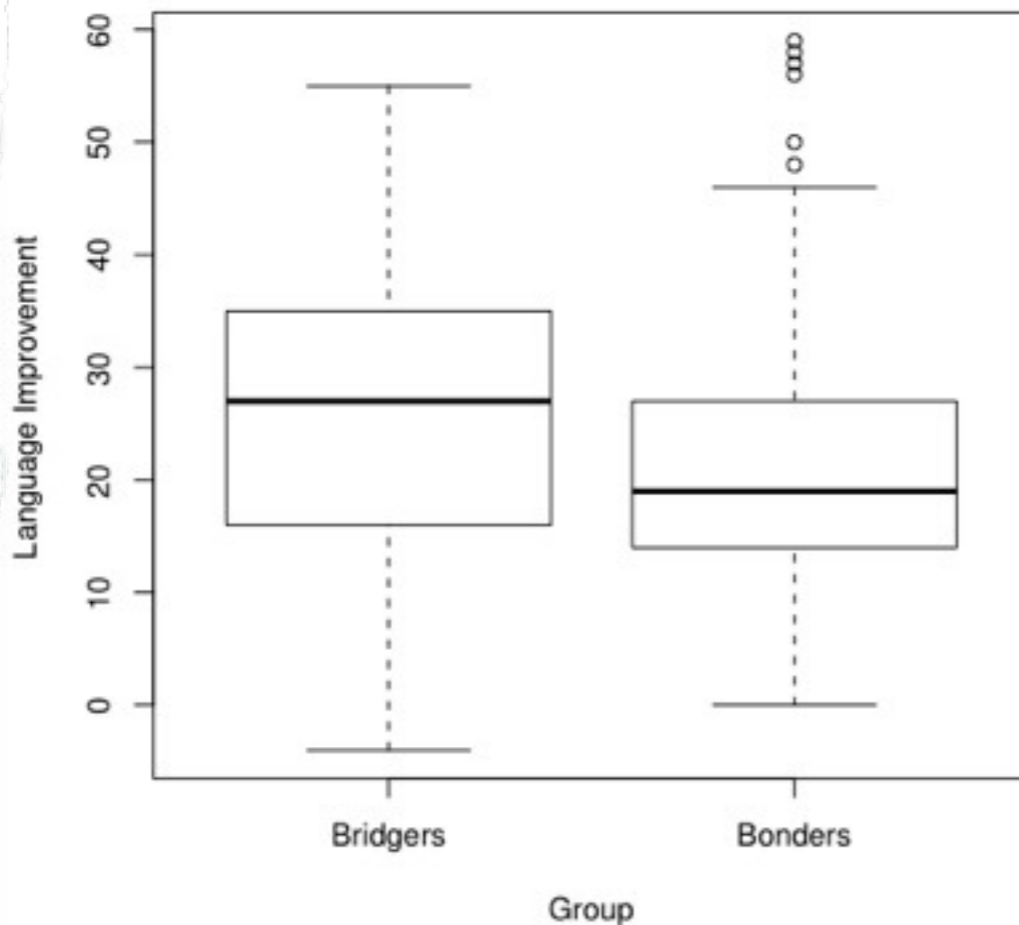
- 📌 Study abroad social interaction questionnaire (SASIQ)
- 📌 Friends and topics about which they spoke were listed
- 📌 Bonding and bridging scores were computed



Results

Table 1: Influence of Social Capital on Conversation:
Bridgers discuss more topics on average than bonders.

Group	Topics	Participants	Topics / Participant
Bonders	786	114	6.89
Bridgers	992	90	11.02



Publications (Case Studies)

- 📌 **Social Capital in the Blogosphere: A Case Study** (published in *Papers from the AAAI Spring Symposium on Social Information Processing*)
- 📌 **Bonding vs. Bridging Social Capital: A Case Study in Twitter** (published in *International Symposium on Social Intelligence and Networking*)
- 📌 **Identifying Health-Related Topics on Twitter** (published at *International Conference on Social Computing, Behavioral-Cultural Modeling, & Prediction*)
- 📌 **Social Capital and Language Acquisition during Study Abroad** (published in *The 33rd Annual Conference of the Cognitive Science Society*)

Conclusion

- 📌 Case Study Contributions
 - 📌 Blogosphere - Highlighted potential bonding/bridging
 - 📌 Twitter - Tested and confirmed principle of homophily
 - 📌 Public Health - Increased understanding of tobacco-related behavior in Twitter
 - 📌 Language Acquisition - Bridging relationships achieve higher levels of language improvement



Conclusion

- 📌 **Computational Framework for Social Capital**
 - 📌 Unifies multiple definitions of social capital into general framework
 - 📌 Uses relationships, affinities, and resources
 - 📌 Hybrid networks - distinguishing potential and realized social capital
 - 📌 Bonding and bridging metrics vary independently (Putnam)
 - 📌 Supports *access* to and *mobilization* of resources
 - 📌 Notion of *interaction* formalized
 - 📌 Offers bridge between resource-sharing sites and social networks

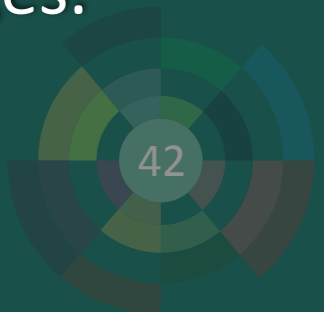


Comparison of SCF to Popular Applications

<i>Feature</i>	FB	TW	LI	FC	SW	SCF
relationships						
simple (friend or not)	✓	✓	✓			✓
complex (strength)						✓
asymmetric		✓				✓
dynamic						✓
affinities						
simple (like or not)	✓		✓		✓	✓
complex (how much)						✓
dynamic						✓
resources						
desired by ego				✓	✓	✓
desired by others				✓		✓
available by ego				✓	✓	✓
...willing to give				✓	✓	✓
...to specific others						✓
available by others				✓	✓	✓
...willing to give				✓	✓	✓
...to ego						✓
dynamic rel. (on exchange)						✓
...update on giving						✓
...update on receiving						✓

Future Work

- 📌 Perform additional resource simulations with multiple types of individuals within the network (e.g., producers and consumers, altruists and free-riders).
- 📌 Map practical selection functions to specific tasks (e.g., finding a job, obtaining support, learning a new skill).
- 📌 Utilize prior social exchange theory to seed reasonable hypotheses to then be empirically tested within the framework. Perhaps, suitable individual-level exchange functions could be developed.
- 📌 Design a range of relationship strength updating functions that allow external feedback (e.g., holidays, weather, crisis) to affect changes.



Future Work

- 📌 Explore the economics of automatically adjusting resource values based on the supply and demand within the network.
- 📌 Test additional well-developed theories of the social sciences to confirm validity within the context of specific online communities (e.g., can bonding social capital on Twitter be leveraged to lose weight?)
- 📌 Observe and track socially connected communities that actively exchange resources.
- 📌 Many of these ideas are finding their way into local startups (e.g., kalood, stubtopia)

