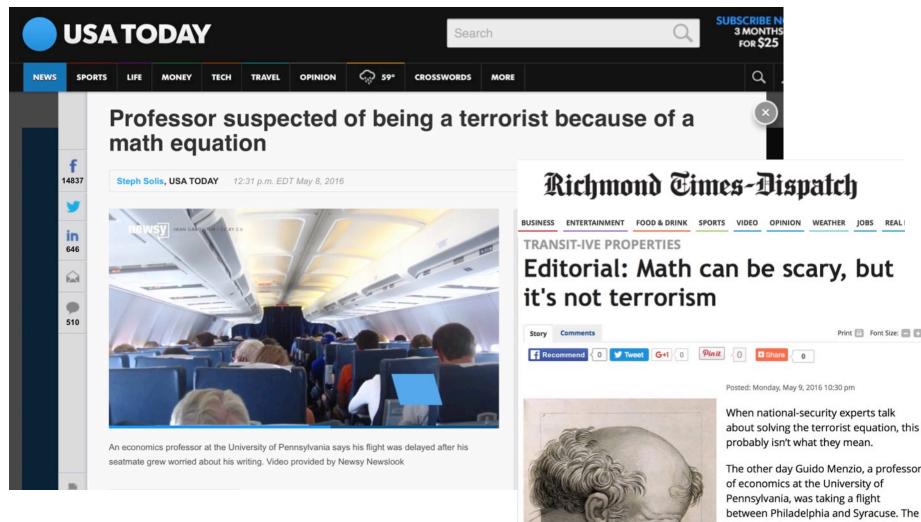
A Relational Practice Approach to the Study of Social Networks

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JOBS REAL

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The other day Guido Menzio, a professor of economics at the University of Pennsylvania, was taking a flight between Philadelphia and Syracuse. The woman in the seat next to him tried to engage him in conversation, but he brushed her off. He was too busy working out a differential equation related to a price-setting model.

But the curious runes he was scribbling were all Greek to his seatmate. Or rather, perhaps all Arabic. She slipped a note to the flight crew, and the next thing Menzio knew, the plane had taxied back to the jetway and he was being

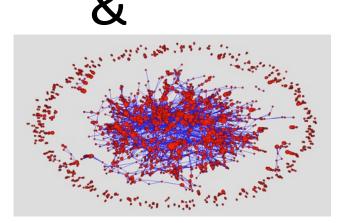
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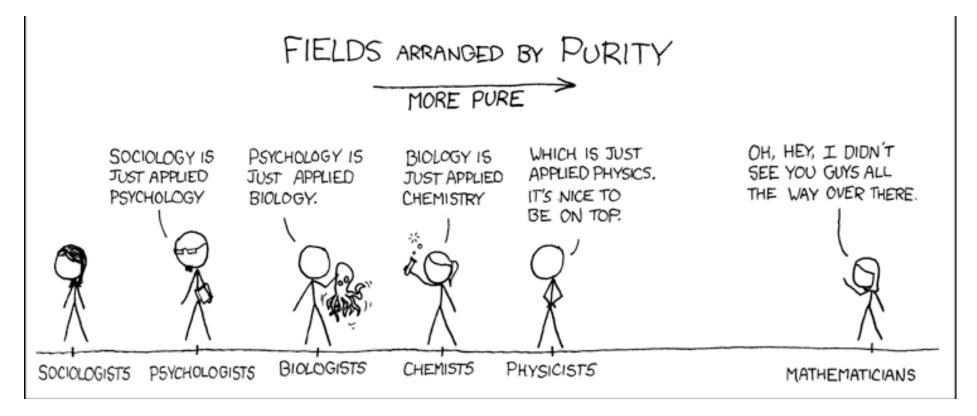






Organizational behavior / Sociology of work

- How do individuals search for and share knowledge to accomplish work?
- How are innovations diffused through knowledge networks within and across bureaucratic organizations?
- How is the changing nature of work affecting work practices?

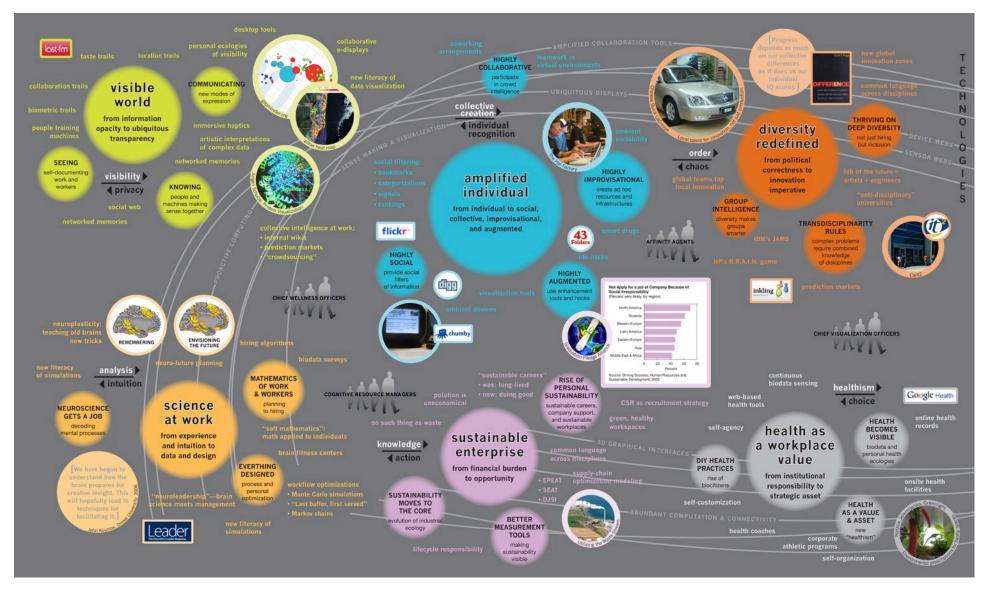


Source: xkcd.com

Research Overview

Торіс	Setting	Findings	Papers	
Conception and implementation of large-scale IT projects	Public officials in state government agencies	 It takes a network to build a network Small and incremental is better than large 	Binz-Scharf 2003, 2008 Binz-Scharf & Lazer 2007 Vaast & Binz-Scharf 2008	
Informal knowledge sharing in highly controlled systems	DNA forensic scientists in government crime labs	 Informal networks arise to compensate for structural deficiencies 	Mergel, Binz-Scharf & Lazer 2008 Binz-Scharf, Lazer & Mergel 2012 Greenberg et al 2012	
Collaborative knowledge production in science	Research scientists in university science labs	 Databases have revolutionized science, but social factors remain strong Relational skills 	Paik & Binz-Scharf 2012, 2014 Binz-Scharf, Kalish & Paik 2015 Binz-Scharf, Dunne & Paik 2016	

Future of Work



Source: Institute for the Future

Studying Organizational Networks

1,200 out of 11,000 papers published in 13 top organization studies journals since 2004 are on social networks (trend rising)*

Social network analysis (SNA) as a quantitative domain (Wasserman & Faust 1994)

Focus on structure (Borgatti, Mehra, Brass, Labianca 2009)

Massive data sets (Lazer, Pentland, Adamic, Aral, Barabasi, Brewer 2009)

What we know less about: content of ties, factors related to tie formation, maintenance, dissolution (Ahuja, Soda, Zaheer 2012; Ghosh & Rosenkopf 2015)

* Of these, 61 papers used some form of qualitative data, 11 were purely qualitative

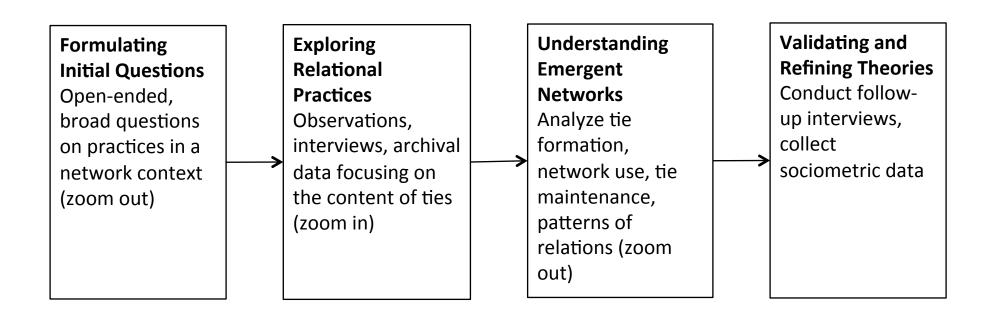
Focus on Practices

Practice is "what is actually done in the doing of work" (Orr 1996:439)

Structures are produced and reproduced as people adopt, adapt, improvise them (Orlikowski 2002)

Networks are something that people *do*, rather than have

Relational Practice Approach



Binz-Scharf, M.C. (2015). A practice approach to the study of social networks. In: Handbook of Qualitative Organizational Research: Innovative Pathways and Methods. Kimberly D. Elsbach and Roderick M. Kramer, eds. New York: Taylor & Francis, pp. 186-194.

An illustration of the relational practice approach

From little science...





...to Big Science

"Collaboration is no longer an option, it's a necessity"

Source: "Sunset of the Solo Scientist", WSJ, Feb 5, 2011

Current Trends in Science

- Co-authored papers more common, team size increasing (Wuchty, Jones and Uzzi 2007)
- Scientific teams span international borders (Olson, Zimmermann and Bos 2008) and institutions (Jones, Wuchty and Uzzi 2008)
- Virtual technologies facilitate knowledge sharing across time and space (Finholt and Olson 1997; Walsh and Roselle 1999)
- In the near future: Open collaboration "Era of Networked Science" (Nielsen 2012)

Democratization of Science

The Challenge

Scientific knowledge production is a collective and social process (Cole 1992; Hacking 1999; Jasanoff 2004; Lynch 1993, Pickering 1992)

Open Science

- Data availability
- Data accessibility
- Drive to understand systems

Sociology

- Organizational boundaries
 - Institutions, schools, departments, labs...
- Socially constructed boundaries
 - Hierarchies, status
 - Trust

Methods

Ethnographic case studies

- Observation of three US labs over 8 months; 50 lab visits total
- Attendance/observation of biennial meeting
- Grounded theory

Social network analysis



- Conference programs of biennial meeting (1996-2012)
- 1,985 papers, 4,971 unique authors
- Coded papers, co-authors, affiliation/location, role (first/last author, chair)
- MRQAP on subset: authors who attended >1; n=219
 - Looked up h-indices, institutional ranking

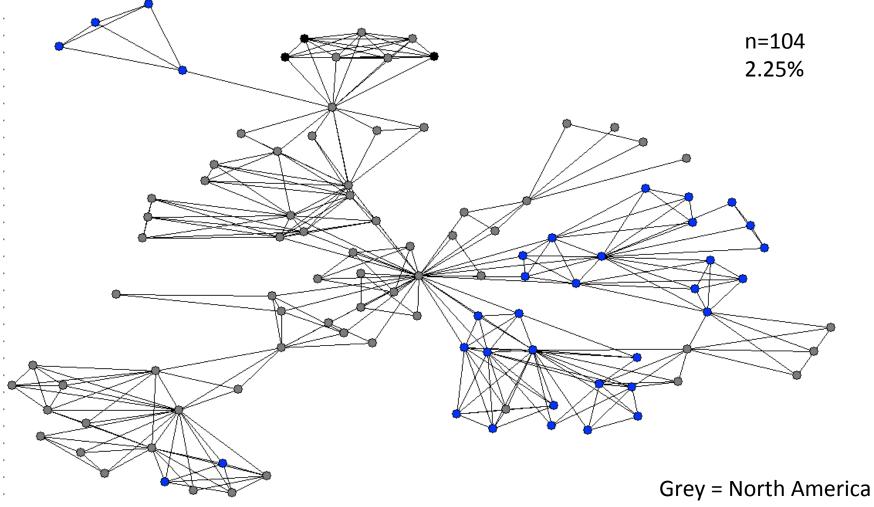
Let's zoom in!

The Way We Work Now

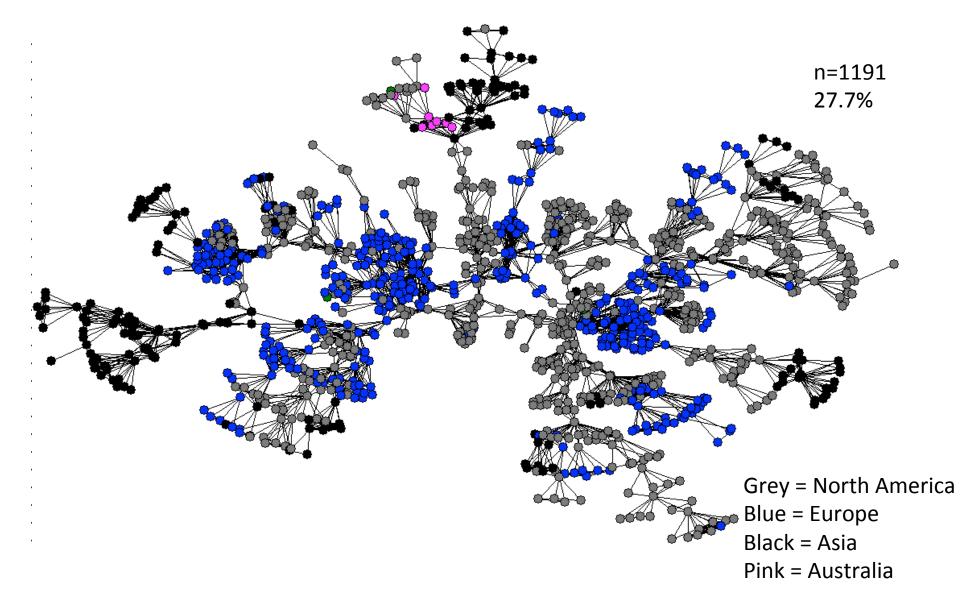
...Based on an ethnography of scientists working in academic labs, we find five categories of relational practices at work: helping, mentoring, apprenticing, co-creating, and **coordinating**. We theorize how these **practices shape scientists'** networks throughout their career stages around [technical and relational] skills, foci, and objects. [...] Our agentic view of network emergence adds to existing understandings of the changing nature of work by emphasizing the **increasingly** entrepreneurial role of knowledge workers in negotiating access to the knowledge they need to accomplish their work. Future research needs to integrate small and big data studies to zoom in and out between relational practices and network structure.

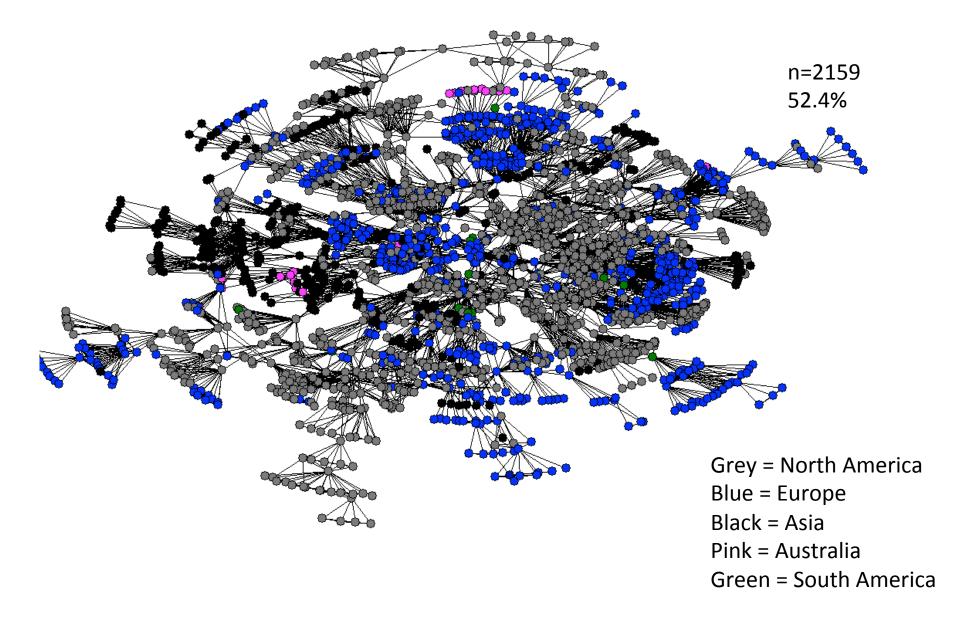
Binz-Scharf, M.C., Danielle D. Dunne, and Leslie Paik (2016). The way we work now: Toward a relational practice theory of network emergence. *Under review*

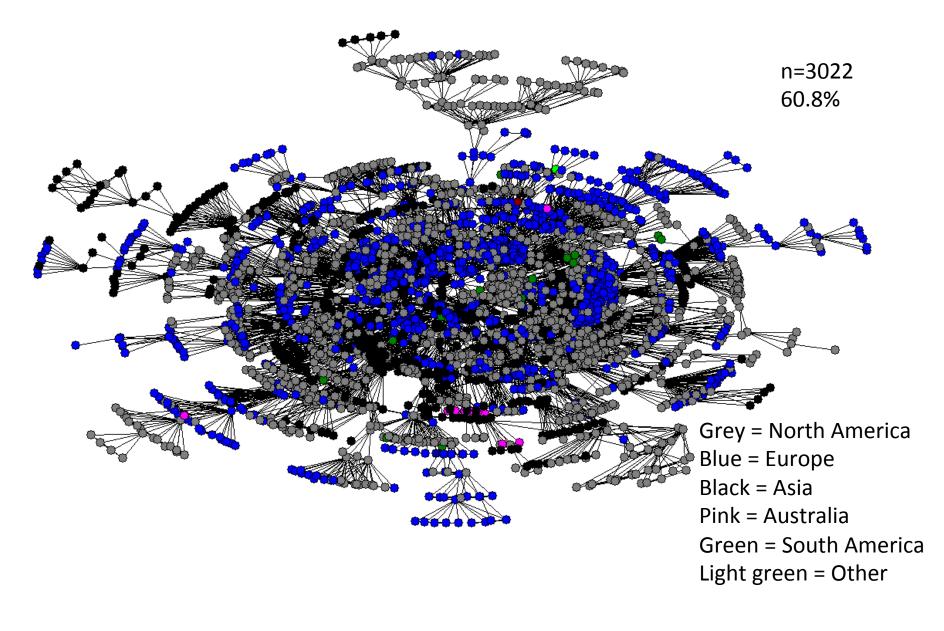
Zooming waaaaay out....



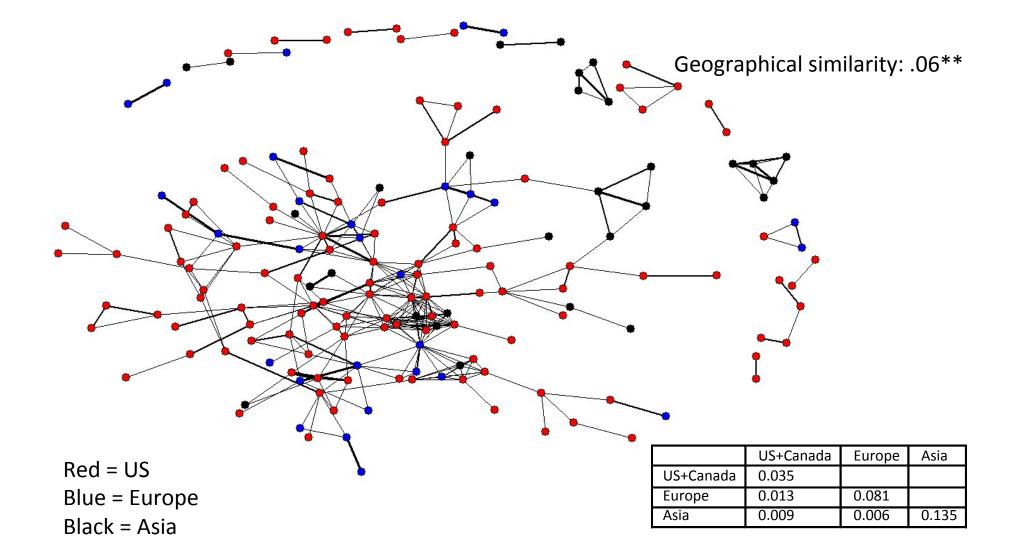
Grey = North America Blue = Europe Black = Asia



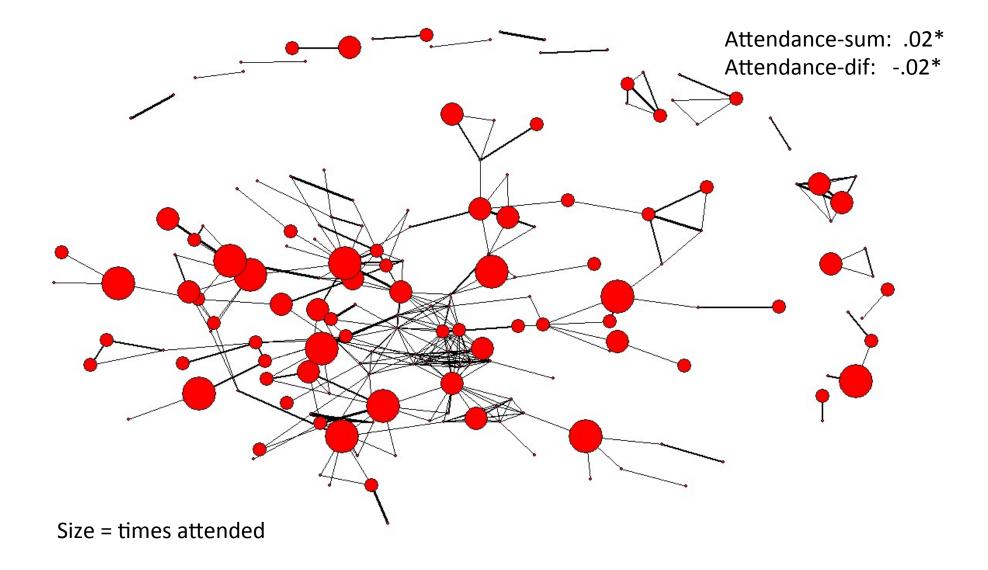




Geographic Regions



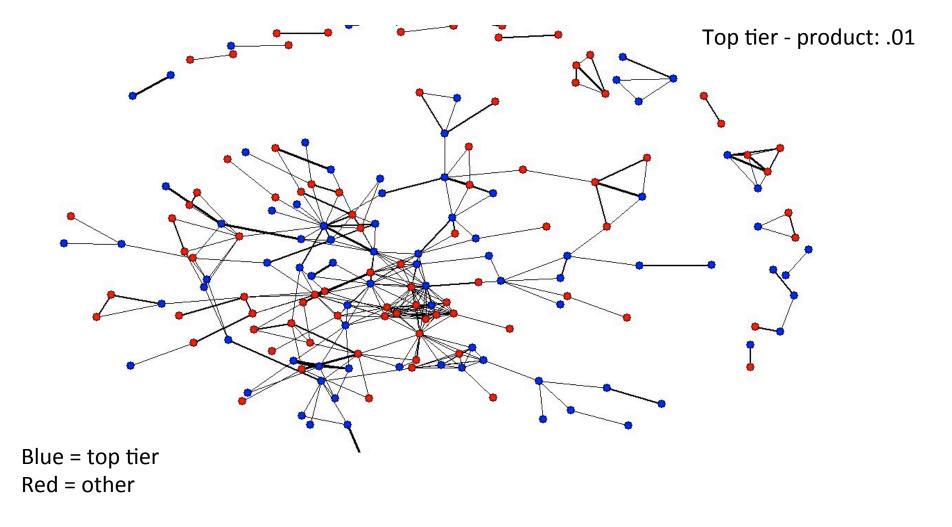
Conference Attendance



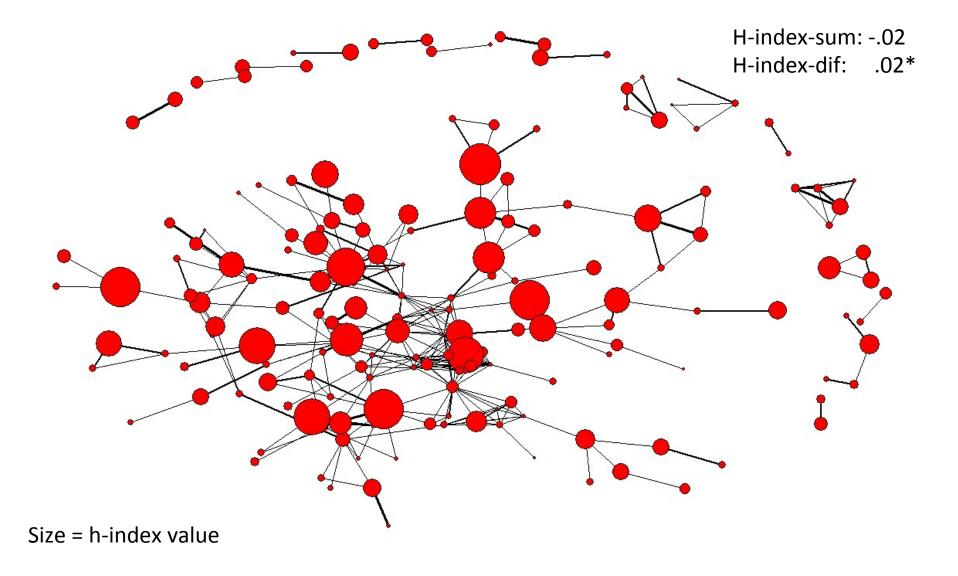
Networking

"For years now... probably 10 years, I see him in meetings, he invited me to give a seminar [at his university]... I would meet him and we would just chat and hang out, and we worked on, sort of, similar things. And then about... a year ago ... we were in a meeting and I said, I've got half a story on this, and he goes, I've got half a story on the same thing. I said come on, let's try and put it together, and see where it goes, and we did, and it was great.."

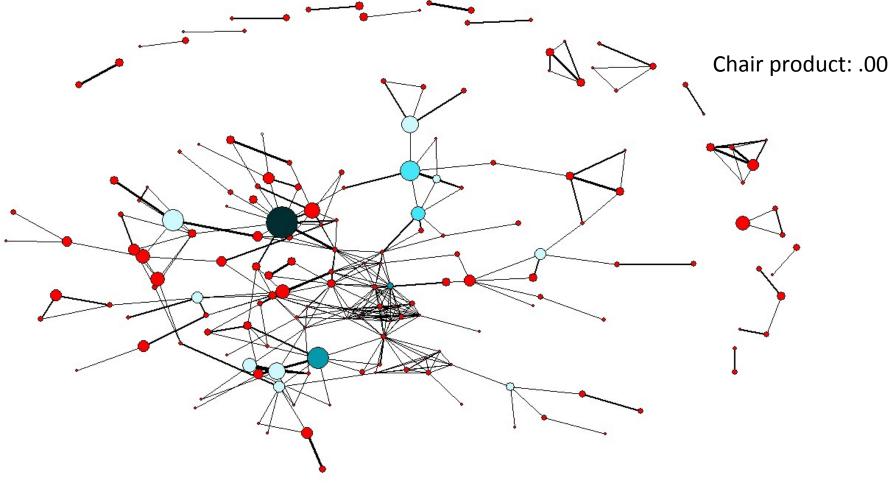
Institutional Ranking



H-Index



Conference Chairs



Blue = chairs (darker attended more) Size = times chaired

Inner Circle

"She's in a different category... Intellectually, she has this capacity to... work to make discoveries that are leaps and bounds ahead of other people. Now at the same time, some of her work seems like really like... generic and boring, and she'll sell it like it's fantastic....But everyone knows that, come on, you know. She'll still get into *Science*."



Competition

Elena [postdoc] says there are some folks who just don't cite Lance's [PI] work even though they do the same things. They don't acknowledge him at all – she says it's weird.

"[On getting a reagent] Now if they are a friend of Marco's [PI], that's not a problem for us. If they're a direct competitor, then we might try to find the reagent from somebody else. [...] [one usually asks what it's for]...Depending on who you're dealing with, they may or may not try to steal your idea." - *Postdoc*

Summary

- Collaboration more likely when institutions are geographically close to one another
- Co-attending conferences leads to collaborations over time
- Institutional ranking has no significant effect on collaboration
- Preferential attachment: High impact scientists attract low impact collaborators
- Pls do not significantly collaborate with one another

Binz-Scharf, M.C., Yuval Kalish, and Leslie Paik (2015). Making Science: Evolution of collaboration in a scientific field. American Behavioral Scientist 59(5):531-547.

Conclusions

- Easier access to data and each other, yet collaborations based on trust relationships
- Traditional structures (roles, hierarchies, departments) remain strong
- Collaborations predicted by face-to-face interactions, preferential attachment, effort to keep competitors at arm's length

Implications

- Focus on individual performance measures for tenure and promotion discourages collaboration
- "Imposing" knowledge sharing doesn't work
- Meetings are important boundary-spanning mechanisms (foci)
- To create a collaborative culture, incentives are needed
 - Revise peer review process
 - Give credit for alternative research outputs, such as shareable resources
- Role models and mentors are key enablers
 - Network grooming

One final implication...

Friend an ethnographer!

Thanks to

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Collaborators

Kyla Bender-Baird Rahul Chhibber Arjun Dhayalan Morgan Hammen Victoria Hill Daniel Lynch Norma Medina Tatiana Ndjatou Jagjit Singh Catherine Vellupilai

P d v

People who did the actual work

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Hypotheses

• <u>Geography</u>: "Death of Distance" (Cairncross 1997) VS. "Distance Matters" (Olson and Olson 2000)

HP 1: Scientists are more likely to collaborate if their institutions are located in the same geographical region

 <u>F2F interactions</u>: Virtual teams function better if reinforced by non-virtual ties (Maznevski&Chudoba 2000); strong ties more conducive to sharing tacit knowledge (Hansen 1999)

HP 2: The more often two scientists attend the same conference together, the more likely they are to collaborate

Hypotheses (cont.)

 <u>Democratizing science</u>: Databases can remove hierarchies of knowledge production (Hine 2006), yet institutional rankings are predictor for collaboration (Jones et al 2008)

HP 3: Two scientists are more likely to collaborate with each other the more similar the rank of their affiliated institution

• <u>Matthew effect</u>: "the rich get richer, and the poor get poorer" (Merton 1968; Jones, Wuchty, and Uzzi 2008); high prestige nodes attract more ties via preferential attachment (Barabasi&Albert 1999)

HP 4: The higher the scientist's prestige, the more likely others collaborate with her/him

HP 5: Scientists collaborate with other scientists of similar levels of prestige

Measures

- Collaboration (dep. variable)
 - Coauthorship on conference paper
- Geography
 - Location based on current affiliation
 - Categorized into regions: US, Europe, Asia
- Conference attendance
 - Number of times attended
- Institutional ranking
 - Academic Ranking of World Universities (ARWU)
 - Six tiers, top-tier categorized as binary variable
- Prestige
 - H-index
 - Paper vs poster presentations
 - Session chair
 - Last author

QAP/MRQAP Results

	В	beta	QAP correlation
Intercept	-0.04	0	
UNI-same	0.01	0.02*	0.02**
GEO-same	0.03	0.06**	0.06**
TOPTIER-product	0.00	0.01	0.03**
ATTEND-sum	0.00	0.02*	0.04**
ATTEND-Diff	0.00	-0.02*	0.02*
APPEAR-sum	0.00	0.12**	0.08**
APPEAR-Diff	0.00	-0.04**	0.04**
Hindex-SUM	0.00	-0.02	0.03**
Hindex-Diff	0.00	0.02*	0.02*
CHAIR-product	0.00	0.00	0.02*
PREGULAR-sum	-0.01	-0.01	-0.01
PREGULAR-diff	-0.04	-0.03**	-0.04**
last_SUM	-0.02	-0.04**	0.01
last_DIF	0.01	0.01\$	0.00
Adjusted R-square	2.4%		

Correlations Between Study Variables

	Coauthorship	APPEAR-	ATTEND-	CHAIR-	GEO-	Hindex-	last_SUM	PREGULAR-	TOPTIER-
	(QAP)	sum	sum	product	same	SUM		sum	product
APPEAR- sum	0.08**								
ATTEND- sum	0.04**	0.52*							
CHAIR- product	0.02*	0.20*	0.10*						
GEO-same	0.06**	-0.02	-0.02	0.01					
Hindex-SUM	0.03**	0.54*	0.22*	0.16*	0.02				
last_SUM	0.01\$	0.41*	0.32*	0.04*	-0.03	0.40*			
PREGULAR- sum	-0.04**	0.10	0.08	0.05*	0.09*	0.34*	0.07		
TOPTIER- product	0.03**	0.09	-0.04	0.05*	0.14*	0.13*	-0.04	0.13*	
UNI-same	0.02	0.02	0.05	0.01	0.10*	0.05	0.04	0.03	0.34*