#### Small-world networks

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#### Outline

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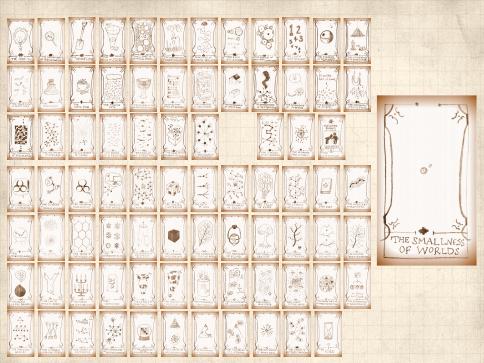
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## People thinking about people:

#### How are social networks structured?

- How do we define and measure connections?
- Methods/issues of self-report and remote sensing.

## What about the dynamics of social networks?

- How do social networks/movements begin & evolve?
- How does collective problem solving work?
- How does information move through social networks?
- Which rules give the best 'game of society?'

#### Sociotechnical phenomena and algorithms:

- What can people and computers do together? (google)
- Use Play + Crunch to solve problems. Which problems?

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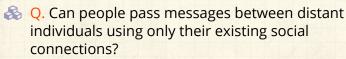


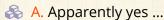


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#### Social Search

## A small slice of the pie:





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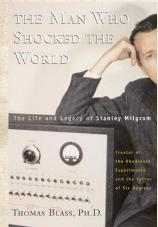






## Milgram's social search experiment (1960s)

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http://www.stanleymilgram.com

- Target person = Boston stockbroker.
- 296 senders from Boston and Omaha.
- 20% of senders reached target.
- $\Leftrightarrow$  chain length  $\simeq$  6.5.

#### Popular terms:

- The Small World Phenomenon;
- "Six Degrees of Separation."



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# From Frigyes Karinthy's "Chain-links" in "Everything is Different", 1929:

'A fascinating game grew out of this discussion. One of us suggested performing the following experiment to prove that the population of the Earth is closer together now than they have ever been before. We should select any person from the 1.5 billion inhabitants of the Earth-anyone, anywhere at all. He bet us that, using no more than five individuals, one of whom is a personal acquaintance, he could contact the selected individual using nothing except the network of personal acquaintances. For example, "Look, you know Mr. X.Y., please ask him to contact his friend Mr. Q.Z., whom he knows, and so forth."

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#### Six Degrees of Kevin Bacon:



♣ The Oracle of Bacon

#### Six Degrees of Paul Erdös:



Academic papers.

🙈 Erdös Number 🗹

🙈 Erdös Number Project 🗹

& So naturally we must have the Erdös-Bacon Number  $oldsymbol{\mathbb{Z}}$ .

 $\red {\$}$  One Story Lab alum has EB#  $< \infty$ .

Natalie Hershlag's (Portman's) EB# = 5 + 2 = 7.

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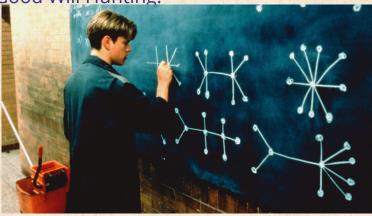
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**Good Will Hunting:** 





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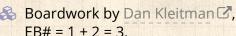
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See Kleitman's sidebar in Mark Saul's Movie Review (Notices of the AMS, Vol. 45, 1998.)





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# You may already be a winner in NSA's "three-degrees" surveillance sweepstakes! NSA's probes could cover hundreds of millions of Americans. Thanks, Kevin Bacon.

by Sean Gallagher - July 18 2013, 4:00pm EDT







Aurich Lawson



A Many people are within three degrees from a random person ...

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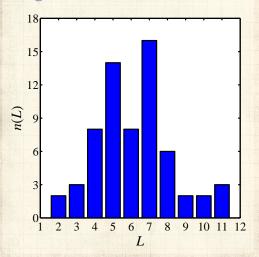




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## The problem

#### Lengths of successful chains:



From Travers and Milgram (1969) in Sociometry: <sup>[12]</sup> "An Experimental Study of the Small World Problem." PoCS, Vol. 1 @pocsvox

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## The problem

#### Two features characterize a social 'Small World':

- 1. Short paths exist, (= Geometric piece) and
- 2. People are good at finding them. (= Algorithmic piece)

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#### Social Search

#### Milgram's small world experiment with email:





"An Experimental study of Search in Global Social Networks"

Dodds, Muhamad, and Watts, Science, 301, 827-829, 2003. [6]

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60,000+ participants in 166 countries



18 targets in 13 countries including

- a professor at an lvy League university,
- an archival inspector in Estonia,
- a technology consultant in India,
- a policeman in Australia, and
- a veterinarian in the Norwegian army.



24.000+ chains

We were lucky and contagious (more later):

"Using E-Mail to Count Connections" , Sarah Milstein, New York Times, Circuits Section (December, 2001)





# All targets:

#### Table S1

Target	City	Country	Occupation	Gender	N	N <sub>c</sub> (%)	r (r <sub>0</sub> )	<l></l>
1	Novosibirsk	Russia	PhD student	F	8234	20(0.24)	64 (76)	4.05
2	New York	USA	Writer	F	6044	31 (0.51)	65 (73)	3.61
3	Bandung	Indonesia	Unemployed	M	8151	0	66 (76)	n/a
4	New York	USA	Journalist	F	5690	44 (0.77)	60 (72)	3.9
5	Ithaca	USA	Professor	M	5855	168 (2.87)	54 (71)	3.84
6	Melbourne	Australia	Travel Consultant	F	5597	20 (0.36)	60 (71)	5.2
7	Bardufoss	Norway	Army veterinarian	M	4343	16 (0.37)	63 (76)	4.25
8	Perth	Australia	Police Officer	M	4485	4 (0.09)	64 (75)	4.5
9	Omaha	USA	Life Insurance	F	4562	2 (0.04)	66 (79)	4.5
			Agent					
10	Welwyn Garden City	UK	Retired	M	6593	1 (0.02)	68 (74)	4
11	Paris	France	Librarian	F	4198	3 (0.07)	65 (75)	5
12	Tallinn	Estonia	Archival Inspector	M	4530	8 (0.18)	63(79)	4
13	Munich	Germany	Journalist	M	4350	32 (0.74)	62 (74)	4.66
14	Split	Croatia	Student	M	6629	0	63 (77)	n/a
15	Gurgaon	India	Technology	M	4510	12 (0.27)	67 (78)	3.67
			Consultant					
16	Managua	Nicaragua	Computer analyst	M	6547	2 (0.03)	68 (78)	5.5
17	Katikati	New Zealand	Potter	M	4091	12 (0.3)	62 (74)	4.33
18	Elderton	USA	Lutheran Pastor	M	4438	9 (0.21)	68 (76)	4.33
Totals				1000	98,847	384 (0.4)	63 (75)	4.05

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Milgram's participation rate was roughly 75%

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Email version: Approximately 37% participation rate.

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Probability of a chain of length 10 getting through:

$$.37^{10} \simeq 5 \times 10^{-5}$$

 $\Rightarrow$  384 completed chains (1.6% of all chains).







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Motivation/Incentives/Perception matter.

If target seems reachable

⇒ participation more likely.

Small changes in attrition rates ⇒ large changes in completion rates

& e.g., \ 15% in attrition rate  $\Rightarrow$   $\nearrow$  800% in completion rate

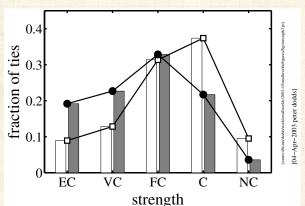






Comparing successful to unsuccessful chains:

Successful chains used relatively weaker ties:



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## Successful chains disproportionately used:

- & Weak ties, Granovetter [7]
- Professional ties (34% vs. 13%)
- Ties originating at work/college
- Target's work (65% vs. 40%)

## ...and disproportionately avoided

- A hubs (8% vs. 1%) (+ no evidence of funnels)
- family/friendship ties (60% vs. 83%)

 $\mathsf{Geography} \to \mathsf{Work}$ 

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Senders of successful messages showed little absolute dependency on

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🚓 age, gender

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country of residence

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income

🚓 religion

relationship to recipient

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Range of completion rates for subpopulations: 30% to 40%





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## Mildly bad for continuing chain:

choosing recipients because "they have lots of friends" or because they will "likely continue the chain."

## Why:



Specificity important



Successful links used relevant information. (e.g. connecting to someone who shares same profession as target.)

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#### Basic results:

- $\langle L \rangle = 4.05$  for all completed chains
- A = Estimated 'true' median chain length (zero attrition)
- A Intra-country chains:  $L_* = 5$
- A Inter-country chains:  $L_* = 7$
- All chains:  $L_* = 7$
- $\clubsuit$  Milgram:  $L_* \simeq 9$

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#### **Usefulness:**

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#### Harnessing social search:

Can distributed social search be used for something big/good?

What about something evil? (Good idea to check.)

What about socio-inspired algorithms for information search? (More later.)

For real social search, we have an incentives problem.

Which kind of influence mechanisms/algorithms would help propagate search?

Fun, money, prestige, ...?

Must be 'non-gameable.'

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## Red balloons:

A Grand Challenge:

A 1969: The Internet is born (the ARPANET —four nodes!).

Originally funded by DARPA who created a grand Network Challenge for the 40th anniversary.

Saturday December 5, 2009: DARPA puts 10 red weather balloons up during the day.

Each 8 foot diameter balloon is anchored to the ground somewhere in the United States.

Challenge: Find the latitude and longitude of each balloon.

Prize: \$40,000.

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\*DARPA = Defense Advanced Research Projects Agency ☑.

#### Where the balloons were:



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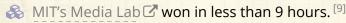






## Finding red balloons:

#### The winning team and strategy:



Pickard et al. "Time-Critical Social Mobilization," [9] Science Magazine, 2011.

People were virally recruited online to help out.

Idea: Want people to both (1) find the balloons, and (2) involve more people.

Recursive incentive structure with exponentially decaying payout:

\$2000 for correctly reporting the coordinates of a balloon.

\$1000 for recruiting a person who finds a balloon.

\$500 for recruiting a person who recruits the balloon finder, ...

(Not a Ponzi scheme.)

🔉 True victory: Colbert interviews Riley Crane 🗹

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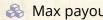






## Finding balloons:

#### Clever scheme:



Max payout = \$4000 per balloon.

Individuals have clear incentives to both

- 1. involve/source more people (spread), and
- 2. find balloons (goal action).
- Gameable?
- Limit to how much money a set of bad actors can extract.

#### Extra notes:

MIT's brand helped greatly.

MIT group first heard about the competition a few days before. Ouch.

A number of other teams did well .

Worthwhile looking at these competing strategies. [9]

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#### Collective Detective:



#### Finding an errant panda

Once again, social media proved to be a powerful dragnet. Around 1:15 p.m., a Washingtonian posted a picture on Twitter of Rusty in a patch of weeds in the Adams Morgan district, not far from the 163-acre zoo, which was created in 1889 by an act of Congress. "Red panda in our neighborhood," wrote Ashley Foughty, who identified herself as a singer, actress and traveler. "Please come save him!"

Another neighbor posted a photograph of two zoo workers, one in safari shorts standing on a rooftop, one holding a giant butterfly net. Soon the zoo announced: "Rusty the red panda has been recovered, crated & is headed safely back to the National Zoo!"



Motherboard, Vice: One Degree of Separation in the Forever War by Brian Castner (November 11, 2015)

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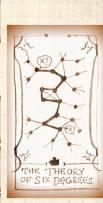
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# The social world appears to be small ...why?

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Theory: how do we understand the small world property?

Connected random networks have short average path lengths:

$$\langle d_{AB} \rangle \sim \log(N)$$

N = population size, $d_{AB}$  = distance between nodes A and B.

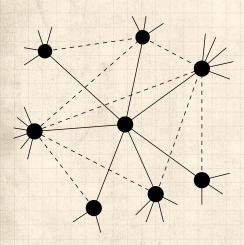








# Simple socialness in a network:



Need "clustering" (your friends are likely to know each other):

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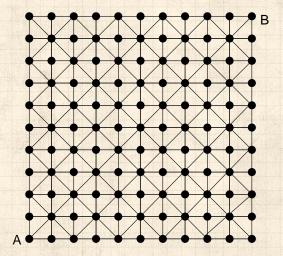
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## Non-randomness gives clustering:



 $d_{AB} = 10 \rightarrow$  too many long paths.

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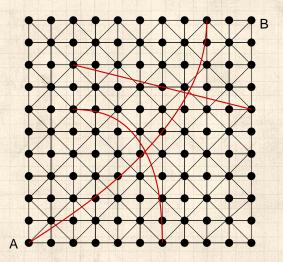
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## Randomness + regularity



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Now have  $d_{AB} = 3$ 

 $\langle d \rangle$  decreases overall





#### Small-world networks

Introduced by Watts and Strogatz (Nature, 1998) [14] "Collective dynamics of 'small-world' networks."

## Small-world networks were found everywhere:

- 🙈 neural network of C. elegans,
- 🗞 semantic networks of languages,
- actor collaboration graph,
- 🚳 food webs,
- 🗞 social networks of comic book characters, ...

## Very weak requirements:

& local regularity + random short cuts

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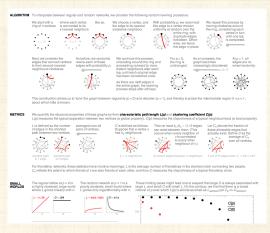
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## Papers should be apps:



Interactive figures and tables = windows into large data sets (empirical or simulated). PoCS, Vol. 1 @pocsvox

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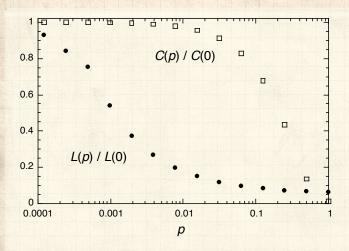
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## The structural small-world property:



& L(p) = average shortest path length as a function of p

& C(p) = average clustring as a function of p

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But are these short cuts findable?

Nope. [8]

Nodes cannot find each other quickly with any local search method.

Need a more sophisticated model ...







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What can a local search method reasonably use?

How to find things without a map?

Need some measure of distance between friends and the target.

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## Some possible knowledge:

Target's identity

Friends' popularity

Friends' identities

Where message has been







Jon Kleinberg (Nature, 2000) [8] "Navigation in a small world."

## Allowed to vary:

- 1. local search algorithm and
- network structure.

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### Kleinberg's Network:

- 1. Start with regular d-dimensional cubic lattice.
- 2. Add local links so nodes know all nodes within a distance q.
- 3. Add *m* short cuts per node.
- 4. Connect i to j with probability

$$p_{ij} \propto x_{ij}^{-\alpha}$$
.

 $\alpha = 0$ : random connections.



 $\alpha$  large: reinforce local connections.



 $\alpha = d$ : connections grow logarithmically in space.



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## Theoretical optimal search:

- "Greedy" algorithm.
- Number of connections grow logarithmically (slowly) in space:  $\alpha = d$ .
- Social golf.

Search time grows slowly with system size (like  $\log^2 N$ ).

But: social networks aren't lattices plus links.

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## Advances for understanding Kleinberg's model:



"Kleinberg Navigation in Fractal Small World Networks"

Roberson and ben-Avraham, Phys. Rev. E, **74**, 017101, 2006. [10]



"Asymptotic behavior of the Kleinberg model" 🗹

Carmi et al., Phys. Rev. Lett., **102**, 238702, 2009. [4]



"Extended navigability of small world networks: Exact results and new insights"

Cartoza and De Los Rios, Phys. Rev. Lett., **2009**, 238703, 2009. [5] PoCS, Vol. 1 @pocsvox Small-world networks

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If networks have hubs can also search well: Adamic et al. (2001)[1]

$$P(k_i) \propto k_i^{-\gamma}$$

where k = degree of node i (number of friends).

Basic idea: get to hubs first (airline networks).



But: hubs in social networks are limited.



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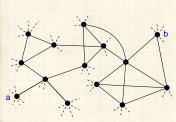






# The problem

If there are no hubs and no underlying lattice, how can search be efficient?



Which friend of a is closest to the target b?

What does 'closest' mean?

What is 'social distance'?

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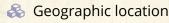


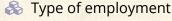
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One approach: incorporate identity.

## Identity is formed from attributes such as:





Religious beliefs

Recreational activities.

Groups are formed by people with at least one similar attribute.

Attributes ⇔ Contexts ⇔ Interactions ⇔ Networks.

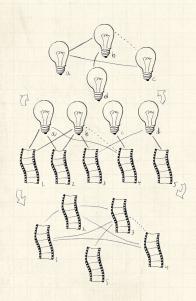
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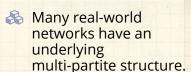




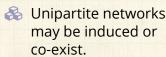


## Bipartite affiliation structures:





- Stories-tropes.
- Boards and directors.
- Films-actors-directors.
- Classes-teachersstudents.
- Upstairsdownstairs.



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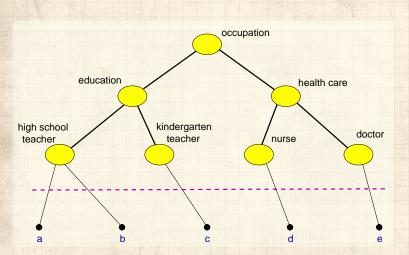
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## Social distance—Context distance



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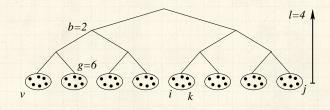
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Distance between two individuals  $x_{i,j}$  is the height of lowest common ancestor.



$$x_{ij} = 3$$
,  $x_{ik} = 1$ ,  $x_{iv} = 4$ .

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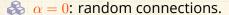
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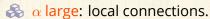
Individuals are more likely to know each other the closer they are within a hierarchy. Small-world
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& Construct z connections for each node using

$$p_{ij} = c \exp\{-\alpha x_{ij}\}.$$





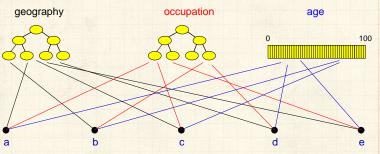






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Blau & Schwartz [2], Simmel [11], Breiger [3], Watts et al. [13]; see also Google+ Circles.

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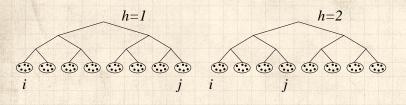
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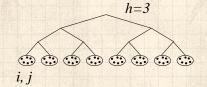






## The model





$$\begin{split} \vec{v}_i &= [1\ 1\ 1]^T \text{, } \vec{v}_j = [8\ 4\ 1]^T \\ x_{ij}^1 &= 4 \text{, } x_{ij}^2 = 3 \text{, } x_{ij}^3 = 1. \end{split}$$

Social distance:  $y_{ij} = \min_{h} x_{ij}^{h}.$ 









### The model

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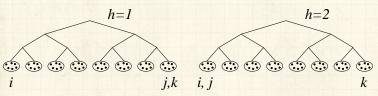
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## Triangle inequality doesn't hold:



$$y_{ik} = 4 > y_{ij} + y_{jk} = 1 + 1 = 2.$$







### The model

Individuals know the identity vectors of

- 1. themselves,
- 2. their friends, and
- 3. the target.

Individuals can estimate the social distance between their friends and the target.

Use a greedy algorithm + allow searches to fail randomly.

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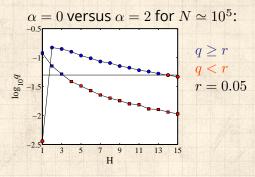
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## The model-results—searchable networks



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q = probability an arbitrary message chain reaches a target.



A few dimensions help.



Searchability decreases as population increases.



Precise form of hierarchy largely doesn't matter.



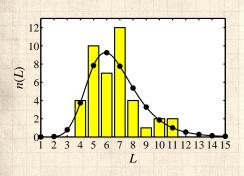




### The model-results

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## Milgram's Nebraska-Boston data:



## Model parameters:

$$\gg z = 300, g = 100,$$

$$b = 10$$
,

$$\alpha = 1, H = 2;$$

$$\langle L_{\rm model} \rangle \simeq 6.7$$

$$A_{\rm data} \simeq 6.5$$

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### Social search—Data

## Adamic and Adar (2003)

- For HP Labs, found probability of connection as function of organization distance well fit by exponential distribution.
- Probability of connection as function of real distance  $\propto 1/r$ .

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### Social Search—Real world uses

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- Tags create identities for objects
- Website tagging: bitly.com
   ✓
- 🙈 (e.g., Wikipedia)
- Photo tagging: flickr.com
- Dynamic creation of metadata plus links between information objects.
- Folksonomy: collaborative creation of metadata

networks Generalized affiliation networks







## Social Search—Real world uses

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### Recommender systems:

- Amazon uses people's actions to build effective connections between books.
- Conflict between 'expert judgments' and tagging of the hoi polloi.







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Nutshell

#### Nutshell for Small-World Networks:

- Bare networks are typically unsearchable.
- Paths are findable if nodes understand how network is formed.
- Importance of identity (interaction contexts).
- Improved social network models.
- Construction of peer-to-peer networks.
- Construction of searchable information databases.







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