

Social Networks and Marriage Matching

Debra Hevenstone

18. November, 2013

Outline

- ▶ Background
 - ▶ Matching
 - ▶ Dating
 - ▶ Hypotheses
- ▶ Method
 - ▶ Simulation Design
 - ▶ Experiment
 - ▶ Results
- ▶ Future work
 - ▶ Simulation
 - ▶ Empirical
 - ▶ Limitations

Matching Problems

- ▶ Goal is a stable match
 - ▶ One:one, one:N
 - ▶ Bipartite 1:1 (called “Marriage Problem”)
- ▶ Algorithms assume
 - ▶ Full scope search
 - ▶ Lengthy search
- ▶ But real marriage search
 - ▶ Limited length search
 - ▶ Limited scope search

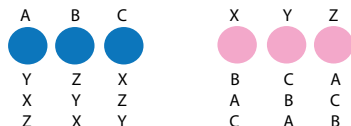
Matching Problems: Algorithms with full scope search

Medical graduates	\longleftrightarrow	Hospitals
H.S. students	\longleftrightarrow	Schools
Organ donors	\longleftrightarrow	Recipients
Rabbinical graduates	\longleftrightarrow	Synagogues
Law school graduates	\longleftrightarrow	Firms



Matching Problems: Gale-Shapely (G-S) algorithm

- ▶ N men and N women rank potential partners (no ties)
- ▶ Men propose to the most-preferred woman not yet proposed to
- ▶ Women accept if unmatched, or if proposal is improvement
- ▶ Up to $n^2 - n + 1$ rounds to stable solution



Stable Solutions

(AZ, BX, CY)

(AY, BZ, CX)

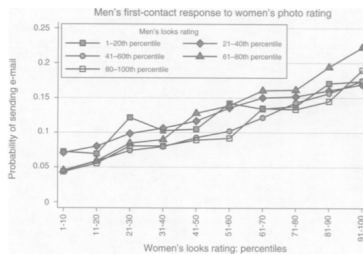
(AX, BY, CZ)

Matching Problems: G-S results & limitations

- ▶ Stable Solution
 - ▶ If a man prefers another, he must have already proposed and been rejected
 - ▶ If a woman prefers another, she must be less preferred than his partner
- ▶ Optimality
 - ▶ Men get best feasible partner
 - ▶ Women get worst feasible partner
- ▶ Real “marriage problem”
 - ▶ Limited time to search
 - ▶ Limited scope for search

Empirical Evidence: Full scope search

- ▶ Full scope: Online dating platform
- ▶ Limited scope: Personal social network
- ▶ G-S algorithm predicting exchange of email/contact info

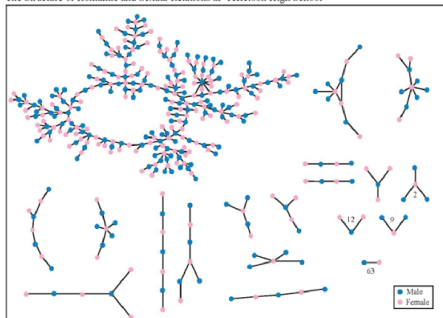


- ▶ Internet dating less homophilous than real life (e.g. edu)

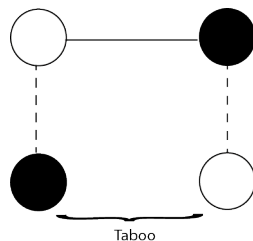
Empirical Evidence: Scope limited by dating history

- ▶ Simulate relationship histories
- ▶ Accurate prediction only limiting scope by relationship history

The Structure of Romantic and Sexual Relations at "Jefferson High School"



Each circle represents a student and lines connecting students represent romantic relations occurring within the 6 months preceding the interview. Numbers under the figure count the number of times that pattern was observed (i.e. we found 63 pairs unconnected to anyone else).



Hypotheses

	<i>Simulation/Mechanism</i>	<i>Empirical Confirmation</i>
<i>Match Utility</i> lower w/ limited scope	Unable to offer good matches	Less happy relationships meeting through friends
<i>Women's Utility</i> higher w/ limited scope lower w/ limited time	Offers to lower-ranked Accepted offers improve with time	Women happier meeting through friends, less meeting young
<i>Dating History</i> loops with longer time/ limited scope, trees with limited time/limited scope	Initial matches follow social network, then cycle w/in clique	Add Health evidence, high school context

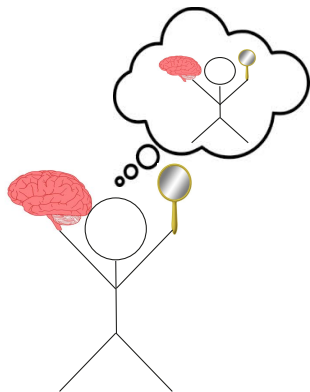
Method: Agent Based Model (ABM)

- ▶ Initialize
 - ▶ Agents w/ heterogenous characteristics & preferences
 - ▶ Calculate match utility & partner rankings
- ▶ Construct social network
 - ▶ Set parameters
 - ▶ Assign edges
- ▶ Run G-S algorithm
 - ▶ Full scope (varied search length)
 - ▶ Scope limited to 2-hop network (varied search length)

Method: Initialization, heterogeneous agents

- ▶ N male agents & N female agents
- ▶ *Symbolic* characteristics & preferences

Attractiveness	$a = \text{norm}(\mu = 5, \sigma = 1)$
Intelligence	$s = \text{norm}(\mu = 5, \sigma = 1)$
Preference for partner's a	$\alpha = \text{uniform}(0, 1)$
Preference for partner's s	$= (1 - \alpha)$



Method: Initialization, match utility & partner rankings

- ▶ Cobb Douglas match utility
- ▶ Constant returns to scale
- ▶ Function of partner's characteristics & ego's preferences

$$u_i = a_j^{\alpha_i} s_j^{1-\alpha_i}$$

u_i	Utility of ego i
a_j	Attractiveness of partner j
α_i	Importance of attractiveness in partner for ego i
s_j	Intelligence of partner j
$1 - \alpha_i$	Importance of intelligence in partner for ego i

Method: Construct network, parameter settings

Parameter	Value
Dating Offers	1
Number of Networks	1
Average Network Degree (should be $<(n-1)/2$)	3
Number of Men	10
Number of Women	10
Coefficient: Impact of attractiveness on making new friends	.03
Coefficient: Impact of intelligence on making new friends	.03
Coefficient: Impact of similar attractiveness on making new friends	-.1
Coefficient: Impact of similar intelligence on making new friends	-.1
Coefficient: Impact of common friends on becoming friends	.1
Coefficient: Impact of already having friends on becoming friends	.1

Network - Matching

Path to where you want to save results: /Users/debrahevenstone/Documents/workspace/

Run Simulation Cancel

Method: Construct network, assign edges

- ▶ Pull two random agents
- ▶ Calculate edge probability

$$Pr_{i:j=1} = \frac{e^f}{1+e^f}$$

$$f = \beta(\mathbf{x}_i + \mathbf{x}_j) + \gamma x_{ij} + \psi(D_i + D_j) + \gamma T_{ij}$$

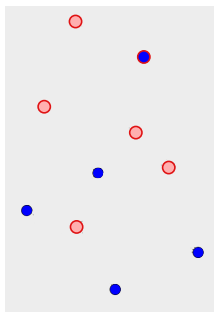
x_i	individual characteristics (<i>not preferences!</i>)
x_{ij}	similarity of characteristics
D_i	current degree
T_{ij}	number of friends in common
$\beta, \gamma, \psi, \gamma$	coefficients

- ▶ if (random uniform (0,1) > $Pr_{i:j=1}$), $(i \cdot j) = 1$

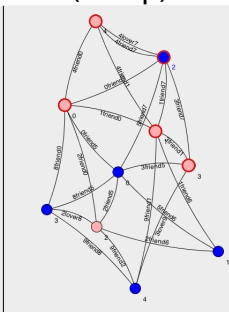
Method: Run G-S Algorithm (full & limited scope)

- ▶ Agents rank *visible* potential partners (no ties)
- ▶ Men propose to the most-preferred *visible* woman (not yet proposed to)
- ▶ Women accept if unmatched, or if better proposal
- ▶ X offer rounds

Full Scope



Limited Scope
(1-hop)



Experiment: G-S matching w/ limited time and scope

- ▶ Hypotheses
 - ▶ Match Utility
 - ▶ Lower w/ limited scope (unable to find best matches)
 - ▶ Female Utility
 - ▶ Higher with limited scope (easier competition)
 - ▶ Lower with limited time (no chance to dump less preferred)

Experiment: Settings

- ▶ Constant conditions
 - ▶ 50 men and women
 - ▶ Agent characteristic distributions ($norm(\mu = 5, \sigma = 1)$)
 - ▶ Coefficients controlling friendship network generation
- ▶ Experimental conditions
 - ▶ Number of offers: 1:10
 - ▶ Full scope search vs limited scope (2-hop) search
- ▶ 10 runs per experimental condition

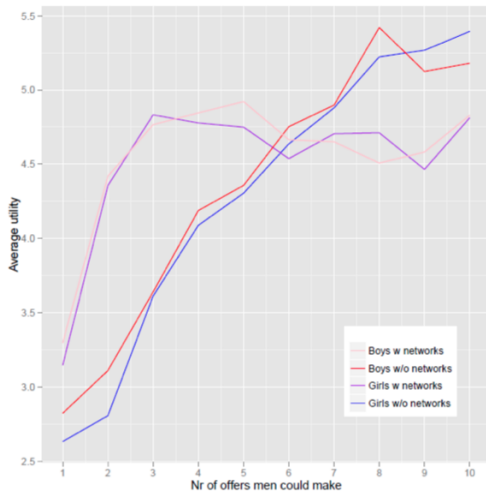
Experiment: Measured output

- ▶ *Agent data*
 - ▶ Individual characteristics
 - ▶ Partner characteristics
 - ▶ **Match utility**
- ▶ Romantic history network (Intermediate steps)
 - ▶ Density at maximum reach
 - ▶ Size of the largest component
 - ▶ Centralization

$$\frac{\sum (centrality_i - centrality_*)}{\max \sum (centrality_i - centrality_*)}$$

- ▶ Mean geodesic length (avg shortest path)
- ▶ Number of 4-cycles

Experiment: Results, match utility



Experiment Results, match utility (OLS)

	coefficient	effect type
<i>Ego variables</i>		
isFemale	-3.68***	built-in
Attractive	1.00***	built-in
Intelligence	1.06***	built-in
<i>Experimental variables</i>		
Number of offers	2.43***	built-in
Number of offers ²	-0.37***	built-in
isNetworkMatch	5.55***	emergent?
<i>Experiment-ego interactions</i>		
isFemale* isNetworkMatch	0.46***	built-in
Attractive *isNetworkMatch	-0.40***	built-in
Intelligence *isNetworkMatch	-0.58***	built-in
isFemale*N. offers	0.46***	built-in
<i>Ego-ego interactions</i>		
isFemale*Attractive	0.38***	emergent?
isFemale*Intelligence	0.51***	emergent?

Also controlling for N. offers * network match

Future Simulation Output: Relationship histories

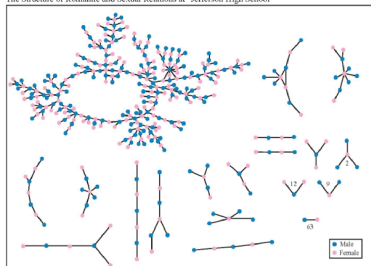
- ▶ Treat intermediate matching steps as relationship history
- ▶ Causal mechanism

limited search scope & limited time

VS

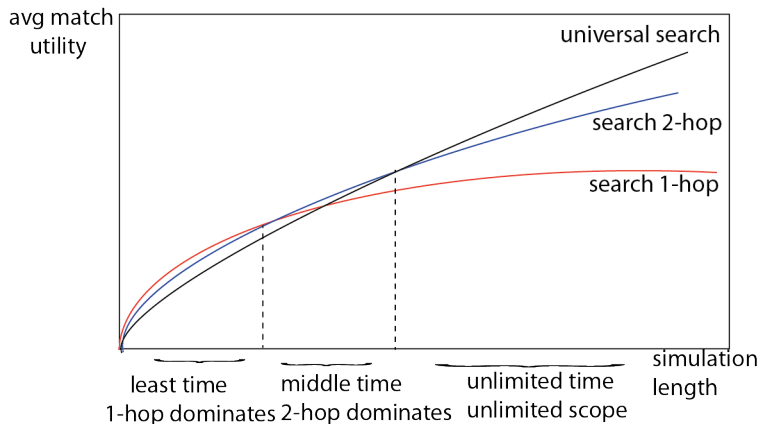
taboos

The Structure of Romantic and Sexual Relations at "Jefferson High School"



Each circle represents a student and lines connecting students represent romantic relations occurring within the 6 months preceding the interview. Numbers under the figure count the number of times that pattern was observed (i.e. we found 63 pairs unconnected to anyone else).

Future Simulation: Continuous information limits



Future Empirical Work: Erasmus/high school dating & friendship networks

▶ Limitations

- ▶ Search time 6 mo to 2 yrs
- ▶ Unlimited search has boundary problems per definition
- ▶ Cross sectional & retrospective

▶ Questions

- ▶ What predicts friendship?
- ▶ Does friendship predict partner?
- ▶ Does partner satisfaction vary if matched over network?
- ▶ Male/female differences?

Empirical Limitations: Timing and causality

- ▶ Study using Facebook data
- ▶ Goal: Identify the relationship edge
- ▶ Two hypotheses
 - ▶ *Embeddedness*
Number of i & j 's friends in common in i 's network
 - ▶ *Dispersion*
Dyads w/ no mutual friend once i and j are removed
- ▶ Causality
Dispersion is observed because edges formed *post*-relationship

Backstrom and Kleinberg, 2013

Conclusion/Who Cares

- ▶ Dating/online dating platforms
- ▶ Epidemiological implications
- ▶ Applications to job search
- ▶ Could imposed networks reduce search time?