## Hierarchy and structure in academic and romantic markets prestige, diversity & inequality

#### Hunter Wapman

PhD Candidate Department of Computer Science hunter.wapman@colorado.edu hne.golf advised by

### Daniel Larremore

University of Colorado Boulder



### faculty shape the academic ecosystem

- make discoveries [science & scholarship]
- teach courses [education]
- train students [research ecosystem & workforce]
- communicate science [media & public]
- advocate for research priorities [policy]

### questions

- 1. faculty production: where do U.S. faculty come from, and are doctoral origins changing?
- 2. placement and prestige: does doctoral institution affect employment? is hiring hierarchical?
- 3. representation of women: is academia heading towards gender parity?

### ten years of comprehensive faculty data

- complete tenure-track faculty rosters
- 10 years (2011-2020) of rosters, collected annually
- all PhD-granting US universities
- all departments, clustered into 107 fields and 8 domains each professor's PhD\* institution & year

\* we treated all doctorates as equivalent

in total: 295,089 faculty in 10,612 departments at 368 universities.





### where do U.S.-trained faculty come from?







### Where do U.S.-trained faculty come from?

over 1 in 8 faculty were trained at just five places:

non-US doctoral programs combined [11%].







### where do U.S.-trained faculty come from?



### did inequalities change from 2011-2020?



#### G are all large across domains.

they do not appear to be growing or shrinking over the decade 2011-2020.



### did inequalities change from 2011-2020?



in every field, domain, and overall, faculty production inequality is lower for new faculty, and higher for sitting faculty!

what might explain these patterns?



# driver: differential attrition risk by PhD origin



this means that there's **substantial inequality in faculty hiring** and that this inequality is then **exacerbated by attrition**.

this process makes cohorts less diverse by doctoral origin as they age.

faculty with the "rarest" PhDs show **nearly 2× the attrition rates** of their colleagues with the most common PhDs.

.

### faculty hiring networks



premises:

- each hiring committee wants to hire the best.
- 2. each hire  $u \rightarrow v$  is an endorsement of u by v.
- network reveals collective mutual endorsements.

faculty hiring figure: Clauset, Arbesman, Larremore. Science Advances, 2015. SpringRank: De Bacco\*, Larremore\*, Moore. Science Advances, 2018.

a recursive notion of prestige:

UC Berkeley

one becomes prestigious when one is endorsed by someone prestigious.

infer prestige scores directly from the structural patterns in faculty hiring networks. [SpringRank — cf. RUMs & Discrete Choice]

convert prestige scores to ranks/percentiles.





### faculty hiring networks

#### low upward mobility

5%↑ Classics
6%↑ Econ, Finance
7%↑ Art History, Stats

12%↑ CS, Epidemiology

18% ↑ Academia
20% ↑ Horticulture
21% ↑ Agronomy, Entomology
23% ↑ Animal Sci, Pathology

#### average hire moves down by

↓28% Econ
↓22% CS
↓18% Academia
↓14% Agronomy
of each field-specific prestige ranking

	ר75.	
ج ج ج	20 20 .5	
e froi	.25-	
ang tot		
ge ch	25	
estiç	5-	
	<u>.</u> .75-	





from 2011-2020:

women's representation **significantly increased** in academia overall, all 8 domains, and 80/107 fields.

it decreased in only 1 field (nursing).



#### women's representation is systematically higher among new hires and lower among attritions in 103/107 fields.

- Academia
- Applied sciences
- Education
- Engineering
- Humanities
- Mathematics and computing
- Medicine and health
- Natural sciences
- Social sciences



#### women's representation is systematically higher among new hires and lower among attritions in 103/107 fields.

demographic curves show why: representation slides downward for cohorts hired in the past.



women's representation among new faculty increased from 2011-2020 overall, in 3/8 domains, and in 15/107 fields (14%)





#### new hires remain predominantly men in 75 of 107 fields, particularly in STEM

women's representation among new faculty increased from 2011-2020 overall, in 3/8 domains, and in 15/107 fields (14%)



#### without continued efforts toward parity in hiring, the changes in women's overall representation from 2011-2020 will soon plateau in many domains.



#### new hires remain predominantly men in 75 of 107 fields, particularly in STEM

women's representation **among new faculty** increased from 2011-2020 overall, in 3/8 domains, and in 15/107 fields (14%)





### summary

- 1. faculty production: a minority of institutions produce the majority of U.S. faculty; attrition exacerbates these inequalities.
- 2. placement and prestige: hiring follows a steep hierarchy at all levels of academia; most faculty work at an institution that is less prestigious than where they earned their doctorate
- 3. representation of women: the share of women in academia is increasing, but slowly and unevenly; without further intervention, progress is likely to stall in most STEM fields

#### explore: Larremore Lab.github.io/us-faculty







### how has dating been studied?

social scientists have long been interested in

- who partners with whom
- why people enter/stay in partnerships
- how current patterns in partnering affect those in the future

the theoretical foundation for most of these studies is the idea that mate pursuit unfolds in a market:

"since men and women compete as they seek mates, a market in marriages can be presumed to exist. Each person tries to find the best mate, subject to the restrictions imposed by market conditions."

s fect those in the future

Gary Becker, "A Theory of Marriage" 1974

### how do market conditions impose restrictions?

you can't "just choose" to partner with the person you like best they might have other suitors, or different preferences.

you can only partner with the person you want if they prefer you over all of their other suitors — only if you successfully beat out the competition.

a suitor can only partner after succeeding in one of the competitions that result from their preferences and opportunities

### assortative mating results from competition



\*there is a rich game-theoretic literature on marriage markets that captures how competition shapes outcomes, but no empirical studies that I know of have looked at competition in relationship markets.

### why hasn't competition been studied?

romantic market and all overtures they made. Such data are only recently available on a large scale (i.e., through online dating).

of transforming those data into games that reveals the competitions.

today I'll present a framework for studying competition and apply it to an online dating market in New York City

- lack of data: studying competition requires information on everyone in a
- lack of methodology: competition isn't directly observable in the data from online dating sites (i.e., who pursues whom). Studying it requires a method

### questions

- 1. competition & popularity: are the most competitive people on the dating site also the most popular?
- 2. who competes with whom: do suitors who pursue similar partners also compete with people who look like them?
- 3. competition stiffness & selectivity: which demographic groups face the "stiffest" competitions (i.e., has the most competitors) and why?

### data from popular, free dating site

- straight singles, 18–65, in New York Metro Area, active on dating site between January - March 2014
- 40.7k men and 36.5k women looking for romantic union 76% white, 82% college educated, average age = 33



#### messaging data

- who contacts whom
- who replies to whom



#### profile data

user demographics (race, age, education)

### from messages to games to competitiveness

#### messaging data

first message	got reply?	
x→a	yes	
y→a	no	mes
z→a	no	pref
x→b	no	
y→b	no	
v→b	yes	hete
w→b	yes	men
y→c	yes	
z→c	no	

#### ...but what if we looked at messaging data in the way people study tournaments and games?

\*Selected Citations: Hirsch, Hortescu, & Ariel 2010; Anderson et al. 2013; Lin & Lundquist 2013; Lewis 2013; Bruch & Newman 2018; Newman & Bruch 2019; Su & Hu 2019; Curington et al. 2020; Curington et al 2021.

saging data have been used to study mate erences & construct desirability rankings\*

rosexual messaging data connect to women & women to men...







### from messages to games to competitiveness

#### messaging data

first message	got reply?
x→a	yes
y→a	no
z→a	no
x→b	no
y→b	no
v→b	yes
w→b	yes
y→c	yes
z→c	no



a suitor wins when an arbiter replies to them, but not to someone else they lose when an arbiter replies to someone else, but not to them

#### games

when a suitor makes an overture to an arbiter, they enter a game

the arbiter's decision to ignore or reply determines the suitor's outcome in the game





### from messages to games to competitiveness

#### messaging data

first message	got reply?
x→a	yes
y→a	no
z→a	no
x→b	no
y→b	no
v→b	yes
w→b	yes
y→c	yes
z→c	no



a suitor wins when an arbiter replies to them, but not to someone else they lose when an arbiter replies to someone else, but not to them

#### win:loss graph games

a suitor with a high win:loss ratio is competitive





### competition through the lens of race

- plays outsized role in shaping people's partner choices, overriding factors such as education
- is the primary focus of previous research on partner  $\bullet$ preferences, especially those using online dating data\*
- is one of the few remaining areas where people openly lacksquareexpress racial preferences

or make generalizations, or as any kind of assertion about worth/value.

note: I'll be using words like "desired, popular, attractive" in their literal sense to make statements about the data and patterns therein — **not** to pass judgements,

Q1: are the most competitive suitors also the most popular?A1: not really.

### competitiveness ≠ popularity



# popularity is consistent with previous findings about preference hierarchies.



see: Ken-Hou Lin and Jennifer Lundquist. Mate selection in cyberspace: The intersection of race, gender, and education.

#### do popularity and win:loss agree?



# do popularity and win:loss *agree*? for white suitors, **yes**



they win and lose in proportion to their popularity.

# do popularity and win:loss *agree*? for non-white men and women, **no**



#### what's going on here?



### lesson: friendship has benefits

competitiveness isn't about attractiveness. it doesn't matter how attractive everyone else thinks you are when you pursue people who disagree.

competitiveness is about getting replies. suitors are most likely to receive replies from arbiters of the same race, so the more often a suitor pursues arbiters of the same race, the more competitive they will typically be.

### Q2: who competes with whom?

### homophily dominates preferences



#### darker = more likely to pursue

Sending Group

Lin, Ken-Hou, and Jennifer Lundquist. 2013. "Mate selection in cyberspace: The intersection of race, gender, and education."



might lead us to expect that suitors mostly compete with members of their own race

### but everyone mostly competes with white competitors...

#### men

		black	latinx	asian	white
race	black	26.2	14.9	5.0	53.8
	latinx	11.8	17.8	5.5	64.9
suitor	asian	6.3	9.4	21.0	63.3
	white	6.0	9.0	5.4	79.6
				_ • _	

% of competitors

women

		black	latinx	asian	white
suitor race	black	24.9	10.6	6.7	57.7
	latinx	10.2	13.5	7.5	68.8
	asian	5.4	6.4	21.1	67.1
	white	4.3	5.5	6.5	83.7

% of competitors

let's focus on asian men, who have a strong preference for asian women.

### how is this possible? follow the messages.





### women's inboxes

#### white women get

- 10% of their overtures from latino men
- 78% of their overtures from white men

#### asian women get

- 24% of their overtures
- from asian men
- 59% of their overtures from white men
- asian men make up only 6% of men on site, so their overtures make up a relatively small fraction of even asian women's inboxes.







make up only 18% of asian men's competition

white men make up fully a third of asian men's competition

women inboxe S

we call this preference dilution



### preferences that affect pursuit are diluted in competition



both positive and negative preferences are diluted.



homophilous messaging doesn't necessarily lead to homophilous competition when the majority of the population comes from one demographic.

we can't understand segregation patterns by only looking at preferences due to this dilution effect.

### lesson: preference dilution

### Q3: who faces the stiffest competitions?



when a suitor pursues an arbiter, the stiffness of that competition is the count of other suitors in pursuit.

the more desirable the arbiter, the stiffer the competition.





#### suitors who enter stiffer competitions are more selective



Strange. Wouldn't we expect most competitive suitors to be in biggest competitions?



### selectivity depends on who you are pursuing



relative selectivity = stiffness when you pursue arbiters of some race - stiffness when others do





### suitors are least selective when pursuing their own race



#### there is a homophily bonus





### suitors are often more selective when pursuing other races



#### there is a *heterogeny tax*

![](_page_51_Figure_4.jpeg)

### suitors are often more selective when pursuing other races

![](_page_52_Figure_1.jpeg)

#### non-white suitors are most selective when pursuing white arbiters

![](_page_52_Picture_3.jpeg)

![](_page_52_Picture_4.jpeg)

### suitors are often more selective when pursuing other races

![](_page_53_Figure_1.jpeg)

#### white suitors apply a heterogeny tax when pursuing non-white arbiters

![](_page_53_Picture_3.jpeg)

![](_page_53_Picture_4.jpeg)

### lesson: social distance is costly

homophily bonus: people evaluate partners who are socially close to them more generously than others do.

heterogeny tax: people evaluate partners who are socially distant from them less generously than others do.

The greater the social distance between suitor and arbiter, the more attractive an arbiter must to be for the suitor to pursue them.

The higher the cost of overcoming social boundaries, the higher the suitor's standards.

### beyond status exchange

#### old idea: people less desirable in one attribute "compensate" with another

e.g., old *but* rich men pursue beautiful but poor women

#### new idea: social distance imposes a cost regardless of suitors' own attractiveness

e.g., black women pursue white men only if they are attractive and educated

#### summary

- 1. competition & popularity: suitors don't need to be desirable to be competitive, as long as they pursue arbiters who are likely to reply
- 2. who competes with whom: preferences have a large impact on who suitors pursue, but a diluted affect on their competition; everyone is most likely to compete with white suitors
- 3. competition stiffness & selectivity: suitors are less selective than their peers when pursuing arbiters of their own race, and they are typically more selective than their peers when pursing arbiters of other races

### reflections and future work

but we really want both types of data in both cases, because then we could ask...

#### in the academic market:

- do men and women apply similarly?
- do their applications succeed at similar rates?
- are there field- or institution-level variations?

 my work on academic markets used placement data my work on romantic markets used application data

#### in romantic markets:

- does homophily increase as pursuit proceeds?
- do market experiences (like desirability or ulletcompetitiveness) affect partnership outcomes?

![](_page_57_Figure_12.jpeg)

#### Thank you to these fantastic people

![](_page_58_Picture_1.jpeg)

Dan Larremore

![](_page_58_Picture_3.jpeg)

Aaron Clauset

![](_page_58_Picture_5.jpeg)

Elizabeth Bruch

![](_page_58_Picture_7.jpeg)

lan van Buskirk

![](_page_58_Picture_9.jpeg)

Katie Spoon

![](_page_58_Picture_11.jpeg)

Allison Morgan

### Hunter Wapman

PhD Candidate Department of Computer Science hunter.wapman@colorado.edu hne.golf

![](_page_58_Picture_15.jpeg)

Sam Zhang

![](_page_58_Picture_17.jpeg)

Nick LaBerge

![](_page_58_Picture_19.jpeg)

Sam F Way

![](_page_58_Picture_21.jpeg)

![](_page_58_Picture_22.jpeg)

![](_page_58_Picture_23.jpeg)

University of Colorado Boulder

![](_page_58_Picture_25.jpeg)