


# How gender shapes our implicit biases about social status: A multi-method approach

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Annenberg School for Communication

February 19, 2020



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# *The Top Jobs Where Women Are Outnumbered by Men Named John*

By CLAIRE CAIN MILLER, KEVIN QUEALY and MARGOT SANGER-KATZ APRIL 24, 2018

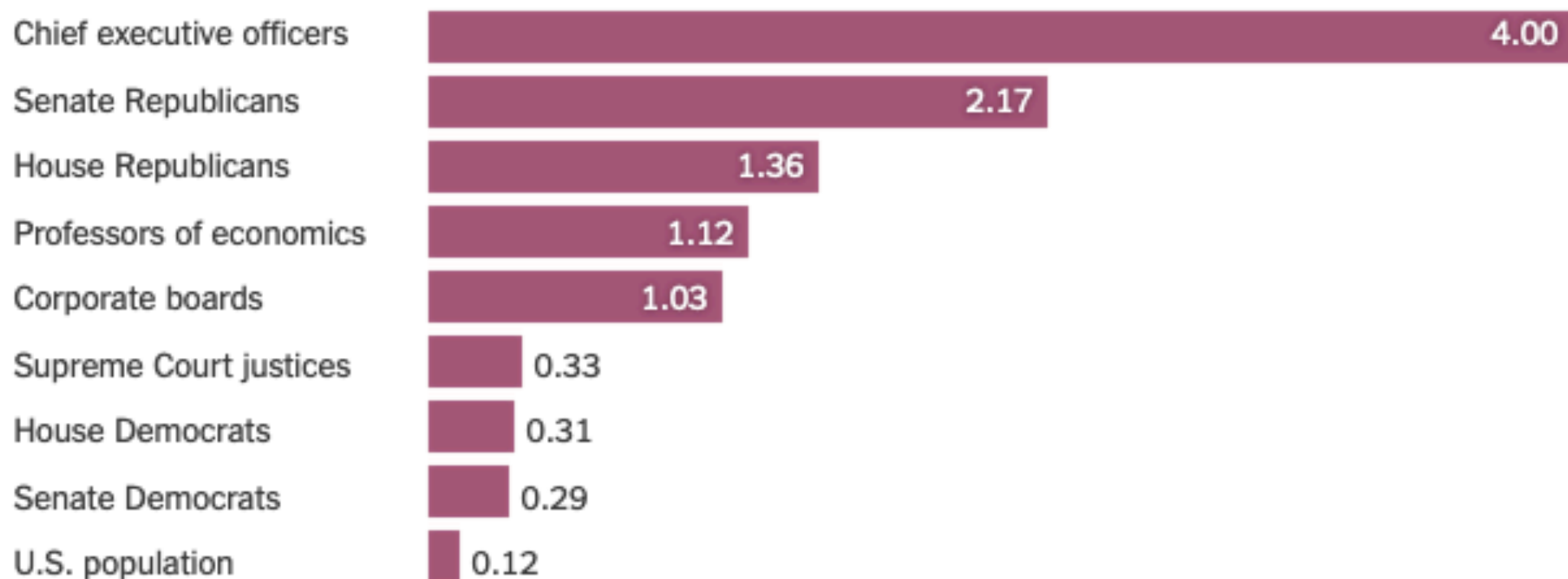
Fortune 500 chief executives named John



# How Thick Is the Glass Ceiling?

We created the Glass Ceiling Index as a measure of the under-representation of women in decision-making roles.

**Ratio of the number of men named James, Robert, John or William to the number of women.**



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Sources: Compustat, Ernst & Young, Roll Call, Center for American Women and Politics, economics department webpages,

# The bigger picture

- Blatant discrimination (Moss-Racusin et al., 2012)
- Stereotype threat (Murphy et al., 2007; Zhang et al., 2009)
- Lack of perceived fit (Cheryan et al., 2017)
- Gender roles shaping identity/values (Eagly, 2009)
- Confidence and effort (Hügelschäfer & Achtziger, 2014; Smith et al., 2013)

# The question

To what extent can gender roles explain differences in implicit preferences for the wealthy in people that vary in gender and socioeconomic rank?

High Status  
=  
Good?

- Mixed evidence in work on explicit status-based evaluations (Cuddy et al., 2008; Varnum, 2013)
- Implicit status-based evaluations are consistently positive
  - Competence (Boukarras et al., 2019)
  - Wealth (Cunningham et al., 2004; Horwitz & Dovidio, 2017; Rudman et al., 2002)
  - Socioeconomic status (SES: Mattan et al., 2019)
- High-status targets tend to elicit greater activity in brain regions involved in positive social evaluations (Mattan et al., 2017, 2018).

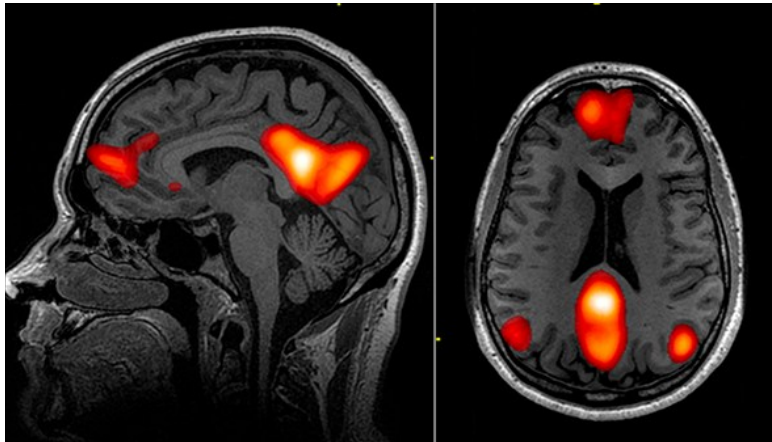
# But what about gender?

- Men compared to women more readily display and pursue status when construed in terms of social or economic influence (e.g., Dovidio et al., 1988; van Vugt & Tyber, 2015).
- Do status-based evaluations differ by gender when not relying on self-report methods?



# Gender identity and precarious manhood

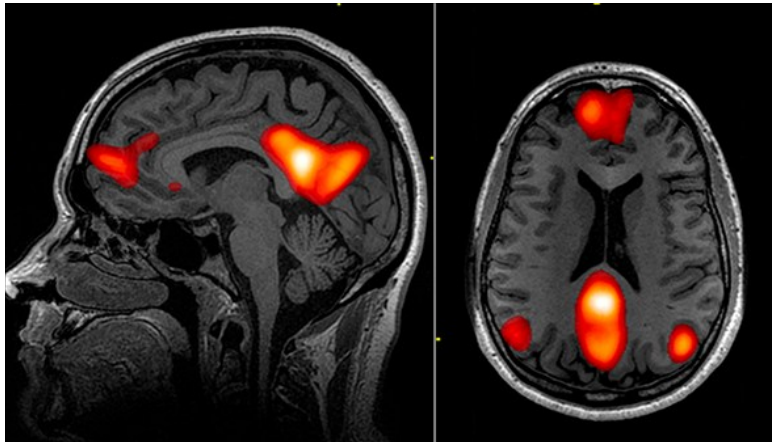
- Unlike femininity, masculinity is seen as a form of status awarded through achievement of cultural standards of manhood and lost through a failure to meet those standards (Gilmore, 1990; Vandello et al., 2008).
- Public reminders of high status (e.g., wealth, professional degrees) could therefore be of greater value to masculine (vs. feminine) identity.



# fMRI: What and why?

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- Measures blood oxygenation levels as an index of metabolic brain activity during a psychological task
- Benefits of fMRI (Mattan et al., 2017)
  - Understanding of neural substrates of motivated impression formation



# fMRI: What and why?

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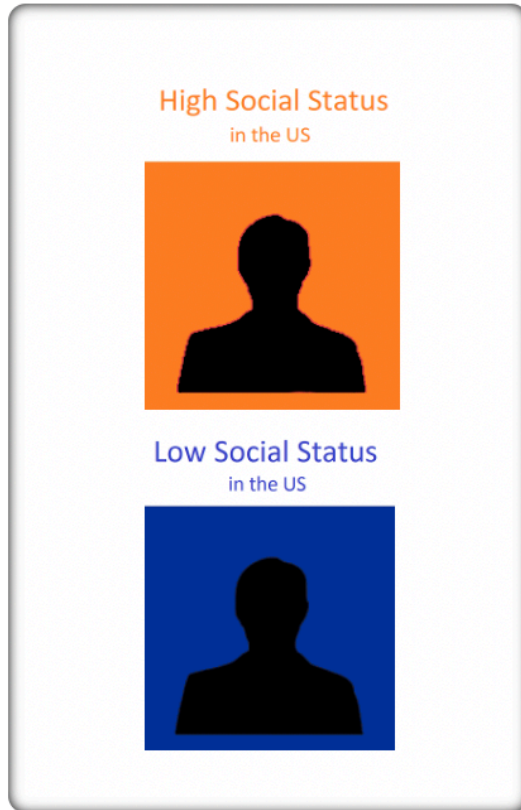
- Measures blood oxygenation levels as an index of metabolic brain activity during a psychological task
- Benefits of fMRI (Mattan et al., 2017)
  - Understanding of neural substrates of motivated impression formation
  - Helps circumvent demand effects

# fMRI Impression Formation Study

- Participants
  - $n=66$  ( $n_{women}=28$ ;  $n_{men}=38$ )
  - White non-Hispanic
  - Lived in the United States for at least 5 years
  - English-speaking
  - Between age 18 and 35
  - No history of drug abuse, psychotropic medication, or diagnosis of developmental or psychiatric disorders.

# Status-color association training

- Participants learned to associate two colors (orange, blue) with low or high socioeconomic status (Mattan et al., 2019).
  - "Those who have the **highest social status** tend to have the most money, the most education, and the most respected jobs."
  - "Those who have the **lowest social status** tend to have the least money, the least education, and the least respected jobs or no job."



# fMRI Impression Formation Study

- Stimuli

- Faces from 14 men and 14 women out of the Kennedy Face Database (Kennedy et al., 2009).
- Equated on attractiveness, likeability, and emotional expressiveness.
- Split equally into two status groupings (low or high), also equated.



1500 ms

+

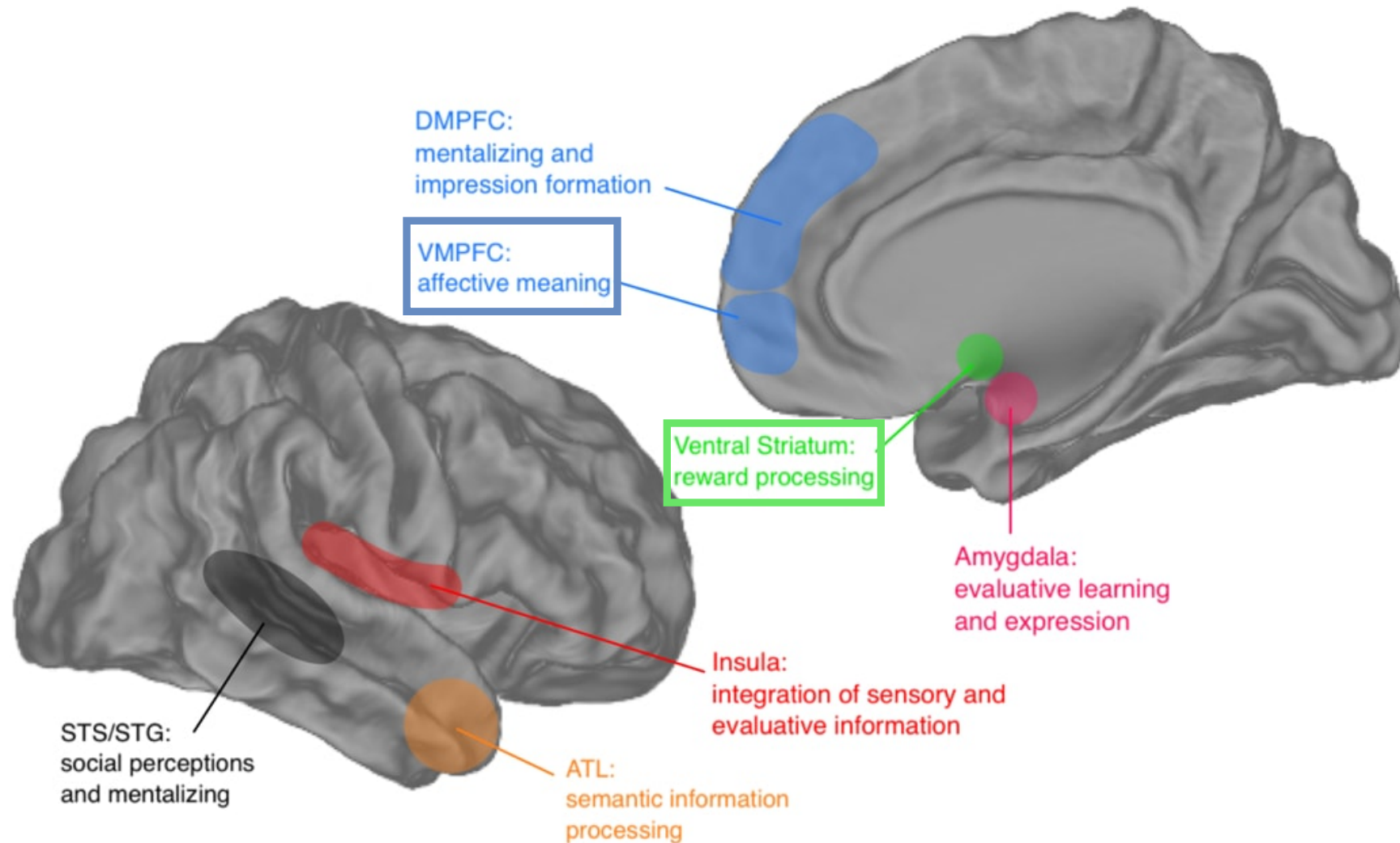
500, 2500, 4500, or 6500 ms





1500 ms

# Neural substrates of status-based evaluations



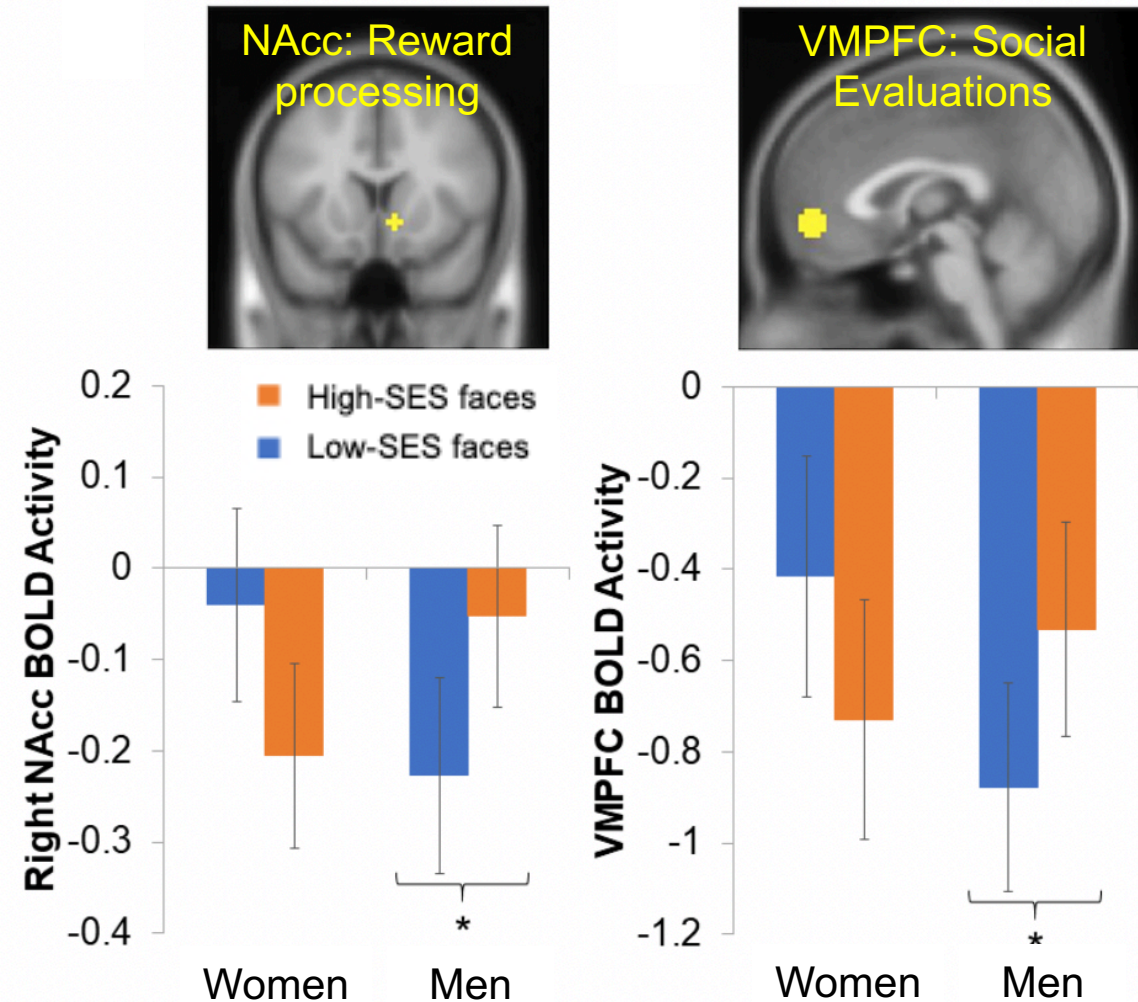
# Predictions for regions supporting positive social evaluations (NAcc, VMPFC)

- Greater response to high status versus low status (Mattan et al., 2017, 2018).
- Larger status effect for men than for women
- Possible effects of face gender
  - Prejudice (main effect)
  - Stereotyping (Gender ✕ Status interaction)
  - Mate Selection

Men show  
greater NAcc  
and VMPFC  
responses to  
high SES

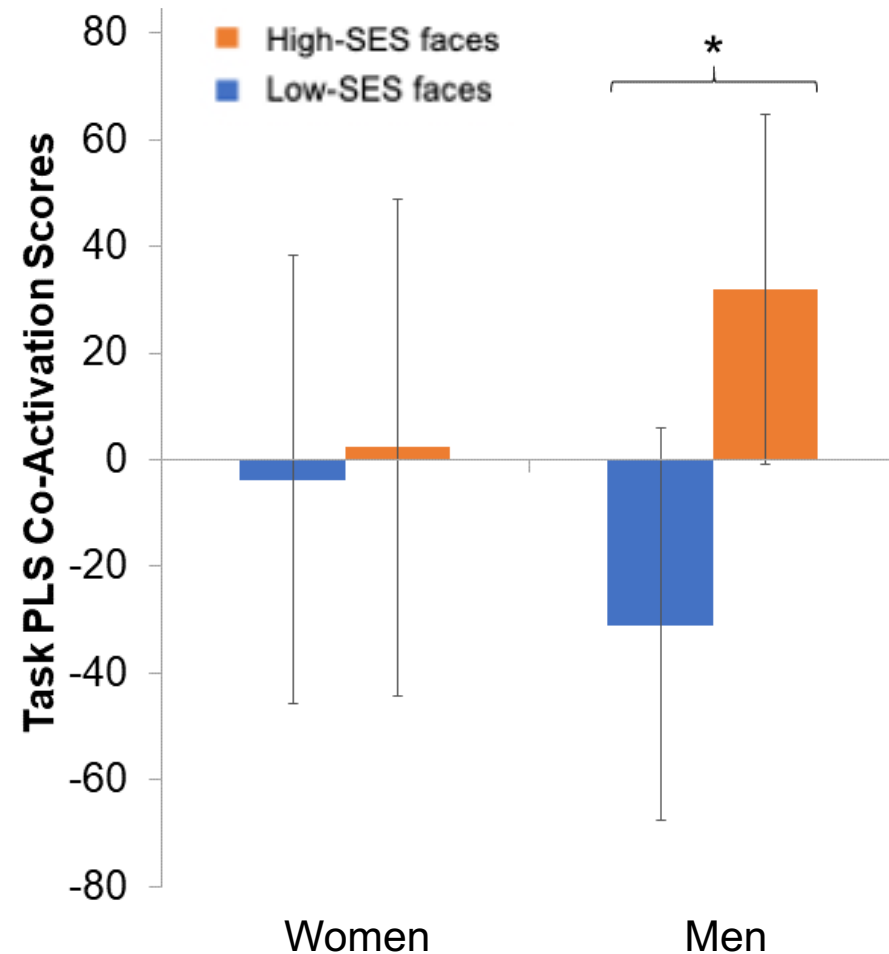
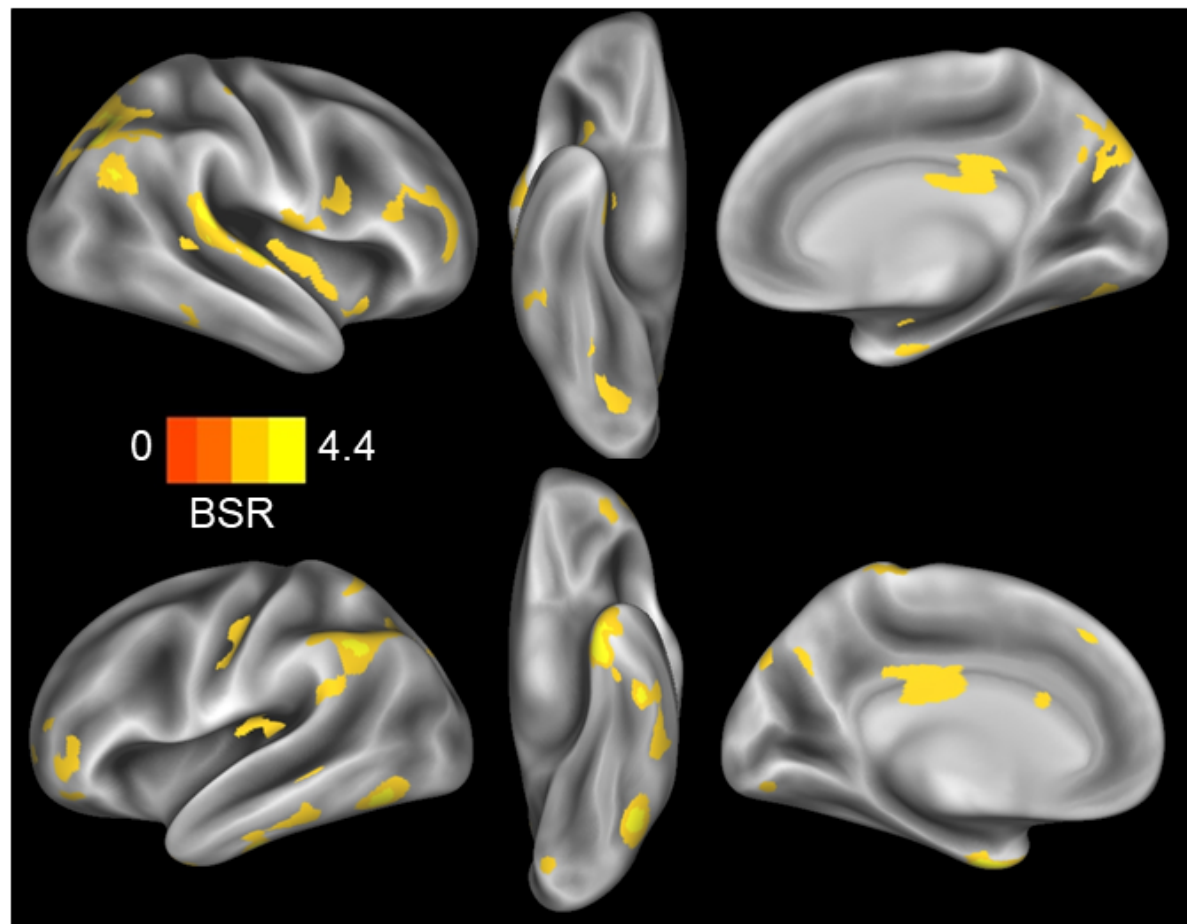
Perceiver Gender  $\times$  Target  
Status Interaction:  $p = .005$ .

Perceiver Gender  $\times$  Target  
Status Interaction:  $p = .008$ .



# Functional Connectivity Analysis (PLS)

# Extended salience/attention network showed greater co-activation based on increasing socioeconomic status



Barth\*, Mattan\*, Dang, & Cloutier, (under review),

\*shared authorship

# fMRI Study: Summary

- Men compared to women show:
  - Greater evidence of preference for high-status faces in regions that support positive social evaluations.
  - Greater coordination among multiple regions involved in salience detection and task engagement.
- No effects of face gender, sexism, or perceiver SES

# Recap

- Do status-based evaluations differ by gender when not relying on self-report methods?
  - Neural level—yes!
  - Implicit associations—maybe?
- Does it also matter how much income or education one has?



# Implicit bias and why it matters

- Reflects activation of valence associations that can serve as input for deliberate evaluations (Gawronski & Bodenhausen, 2011)
- May be particularly important for ambiguous contexts requiring relatively rapid responses (Horwitz & Dovidio, 2017; Moore-Berg et al., 2017)

# Theoretical motivation

- Gender roles (status incongruity)
- Gender identity (masculinity)

# Gender roles and status incongruity

- Women (and men) who violate gender roles frequently face negative backlash (Eagly & Karau, 2002; Moss-Racusin et al., 2010; Rudman, 1998; Rudman et al., 2012).
- Such backlash may trigger ambivalence about high status in status-incongruent individuals (Mattan et al., 2019).
- Previous work suggests that competing (i.e., ambivalent) associations can result in attenuated implicit bias (Petty, Briñol, Tormala, & Jarvis, 2006).

# Gender identity and precarious manhood

- Status gain/loss is important to masculine identity (Gilmore, 1990; Vandello et al., 2008).
- Men weigh high status more heavily when forming impressions of others (Barth, Mattan, et al., under review).
- Income versus education?

# Predicting implicit pro-rich bias

- H1: Income will increase implicit pro-rich bias more for men than for women
- H2: Income and education will both modulate men's (vs. women's) pro-rich bias
  - At low income (H2A): Gender  $\times$  Education
  - At high income (H2B): Education main effect?

Where and how did I find the data?

# The preregistration revolution

**Brian A. Nosek<sup>a,b,1</sup>, Charles R. Ebersole<sup>b</sup>, Alexander C. DeHaven<sup>a</sup>, and David T. Mellor<sup>a</sup>**

<sup>a</sup>Center for Open Science, Charlottesville, VA 22903; and <sup>b</sup>Department of Psychology, University of Virginia, Charlottesville, VA 22904

Edited by Richard M. Shiffrin, Indiana University, Bloomington, IN, and approved August 28, 2017 (received for review June 15, 2017)

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## The Attitudes, Identities, and Individual Differences (AIID) Study and Dataset

Contributors: [Ian Hussey](#), [Sean Hughes](#), [Calvin K. Lai](#), [Charles R. Ebersole](#), [Jordan Axt](#), [Brian A. Nosek](#)

Affiliated institutions: [University of Virginia](#), [Universiteit Gent](#)

Date created: 2012-05-31 01:58 PM | Last Updated: 2019-07-17 07:02 AM

Identifier: DOI 10.17605/OSF.IO/PCJWF

Category:  Project

Description: A large dataset for investigating relations among implicit and explicit attitudes and identities, motivations, beliefs, knowledge, and popular individual difference measures.

## AIID sample characteristics

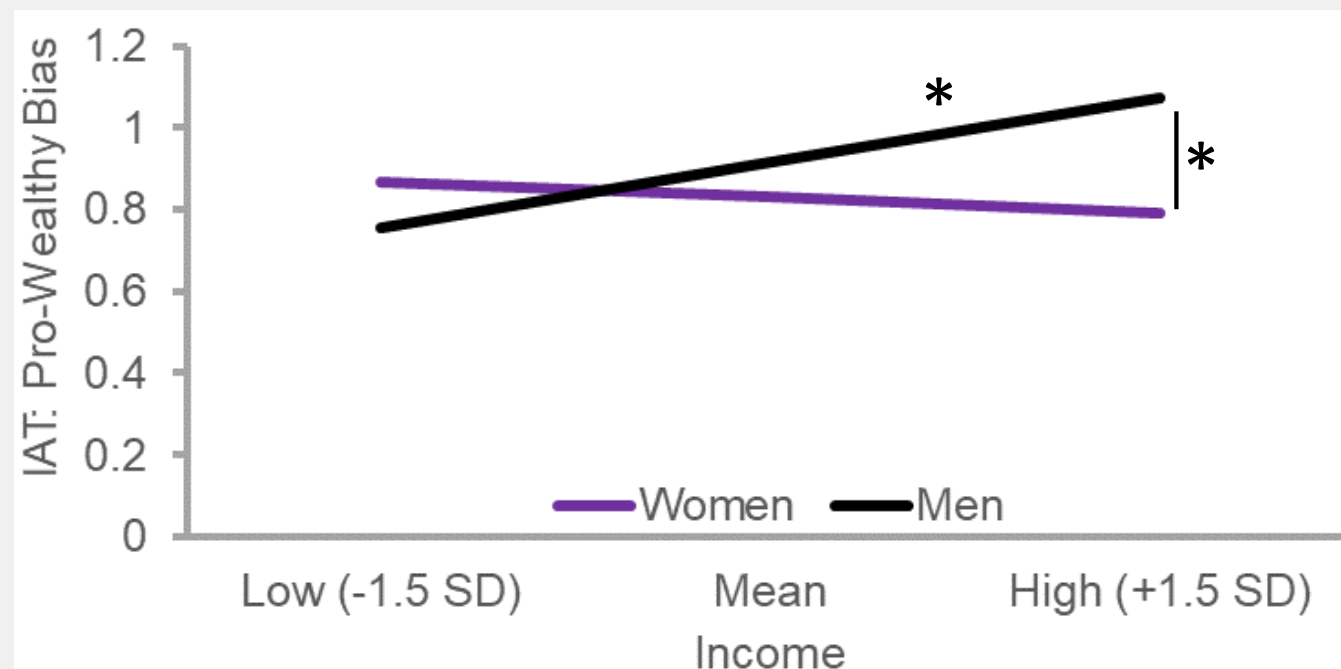
- Data collected online between 2004 and 2007
- Sample size:  $n = 175$  ( $n_{women}=115$ ;  $n_{men}=60$ )
- 71.4% White, 26.3% racial/ethnic minority
- Between age 14 and 66



# AIID measures

- Outcome measure: IAT D score—  
preference for rich over poor
- Predictors:
  - Sex: only options were male and female
  - Income: 5-point scale
  - Education: 5-point scale
- Model:
  - $IAT \sim \text{sex} * \text{income} * \text{education}$

H1: Men  
show greater  
pro-rich bias  
with  
increasing  
income



Gender × Income interaction:

$b = 0.132$ ,  $SE = 0.055$ ,  $CI_{95\%} = [0.024, 0.240]$ ,  $t(167) = 2.413$ ,  $p = .017$

## H2: Income and education modulate gendered preference for the rich

Gender × Income × Education interaction:

$b = -0.089$ ,  $SE = 0.059$ ,  $CI_{95\%} = [-0.205, 0.027]$ ,  $t(167) = -1.521$ ,  $p = .130$

## H2: Income and education modulate gendered preference for the rich

Gender  $\times$  Income  $\times$  Education interaction:

$b = -0.089$ ,  $SE = 0.059$ ,  $CI_{95\%} = [-0.205, 0.027]$ ,  $t(167) = -1.521$ ,  $p = .130$

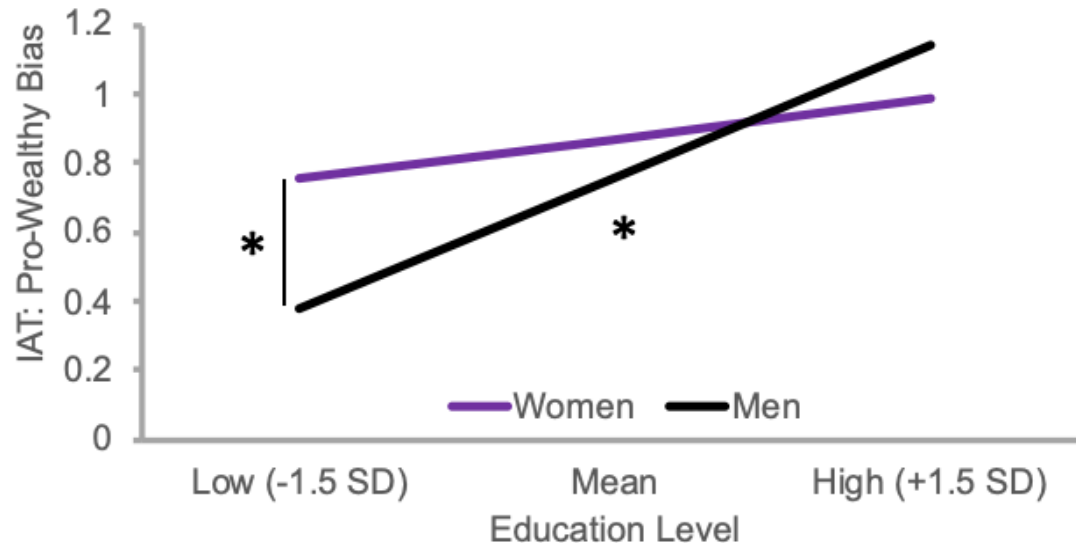
Gender  $\times$  Education interaction:

$b = 0.180$ ,  $SE = 0.113$ ,  $CI_{95\%} = [-0.043, 0.403]$ ,  $t(167) = 1.591$ ,  $p = .113$ .

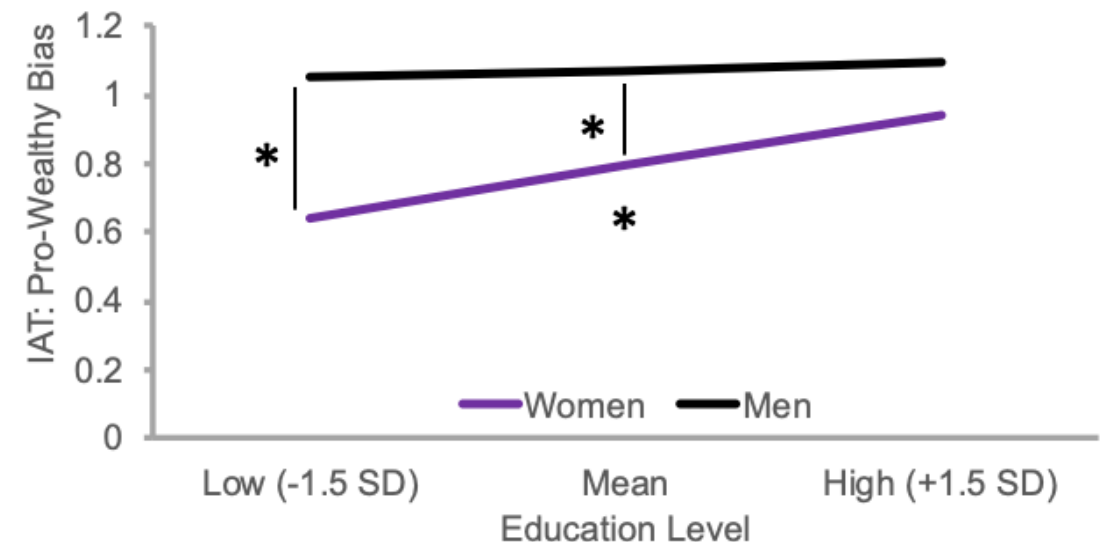
Gender  $\times$  Education interaction:

$b = -0.089$ ,  $SE = 0.097$ ,  $CI_{95\%} = [-0.279, 0.102]$ ,  $t(167) = -0.917$ ,  $p = .360$ .

Low Income Levels (-1.5 SD)



High Income Levels (+1.5 SD)



# IAT Study: Summary

- Men compared to women show greater implicit bias in favor of the rich over the poor as a function of:
  - increasing income, and
  - increasing education levels (low income only)
- No effects of income for women

# Discussion and Final Takeaways

- Across multiple methods, men evaluate high status more positively than women
- Results partly consistent with gender roles and masculine identity

# Discussion and Final Takeaways

- Across multiple methods, men evaluate high status more positively than women
- Results partly consistent with gender roles and masculine identity
- Findings challenge generalizations from the mate selection literature

# Thank You!



Impression Formation Social Neuroscience Lab (University of Delaware):

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Denise Barth

Tzipporah Dang

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Emily Falk

Disparities and Diversity Science Group

Brian Nosek and colleagues at Project Implicit and OSF

Michael Kraus and Anonymous Reviewers



# Registered report pilot sample: Distribution by gender and SES

*Table 1. Distribution of participants from the pilot dataset (n = 175) by gender, income level, and education level.*

Gender	Income Level (USD)	No High School Diploma	High School Graduate	Associate's Degree or Some College	Bachelor's Degree	Graduate Degree or Education
Women	< \$25,000	0	0	12	6	1
	\$25,000 - \$49,999	0	5	8	17	3
	\$50,000 - \$74,999	0	1	7	10	8
	\$75,000 - \$149,999	0	2	7	8	7
	> \$150,000	1	1	2	4	5
Men	< \$25,000	0	0	8	7	1
	\$25,000 - \$49,999	1	0	2	6	0
	\$50,000 - \$74,999	0	2	3	4	4
	\$75,000 - \$149,999	0	0	4	9	5
	> \$150,000	0	0	1	0	3

*Note.* Numbers within each cell indicate the sum total of pilot participants in that condition.

# Registered report: Participant-level exclusions $n_{pilot} = 46$

(1) Greater than or equal to 35% of responses under 300 ms in any one practice block.

(2) Greater than or equal to 25% of responses under 300 ms in any one critical block.

(3) Greater than or equal to 10% of responses under 300 ms in critical blocks.

(4) Greater than or equal to a 50% error rate in any one practice block.

(5) Greater than or equal to a 40% error rate in practice blocks.

(6) Greater than or equal to a 40% error rate in any one critical block.

(7) Greater than or equal to a 30% error rate in critical blocks.

(8) In addition to implementing criteria 1–7 that were used by Nosek and colleagues (2007), we also adopted a stricter exclusion criterion, removing any participant with greater than or equal to 10% responses over 10,000 ms in IAT critical blocks. This additional criterion was implemented to exclude participants who were potentially insufficiently attentive during, or confused by, the IAT.