

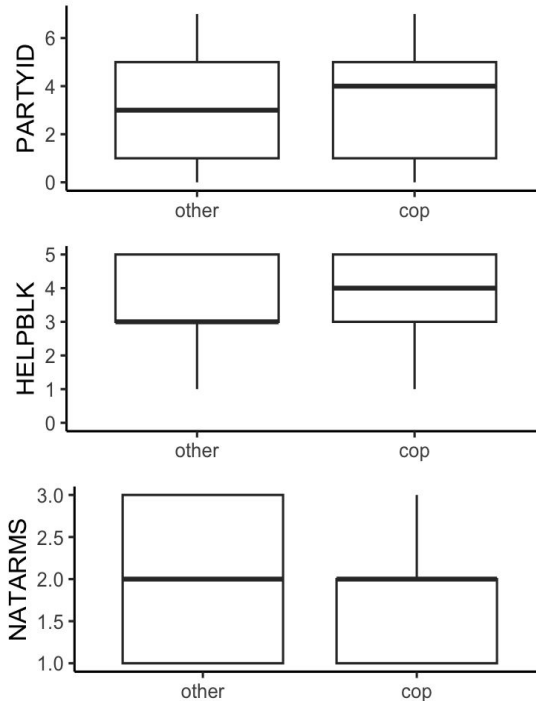
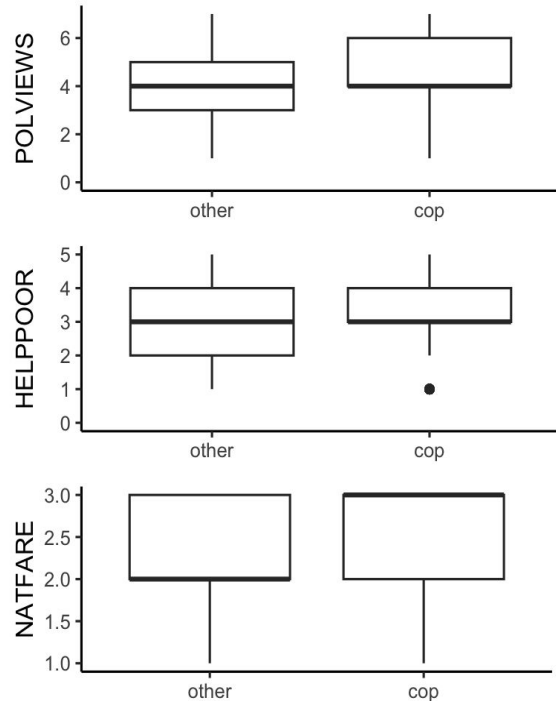
# THE POLITICAL LEANINGS OF POLICE

STAT 3105 – Final Project

Marlon Kegel

# THE QUESTION

(American) police officers political views differ from those of people in other professions.



“[67%] of all police officers say the deaths of blacks at the hands of police are isolated incidents” vs. “only about four-in-ten members of the public (39%) share this view”.

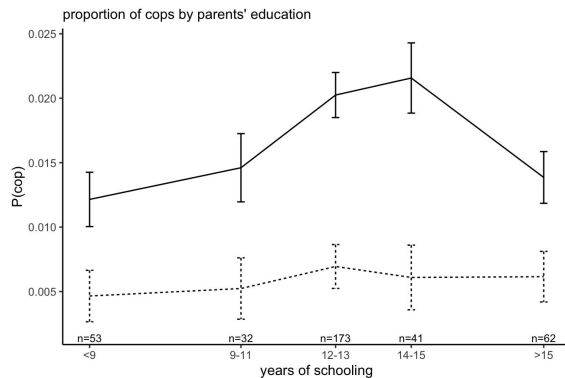
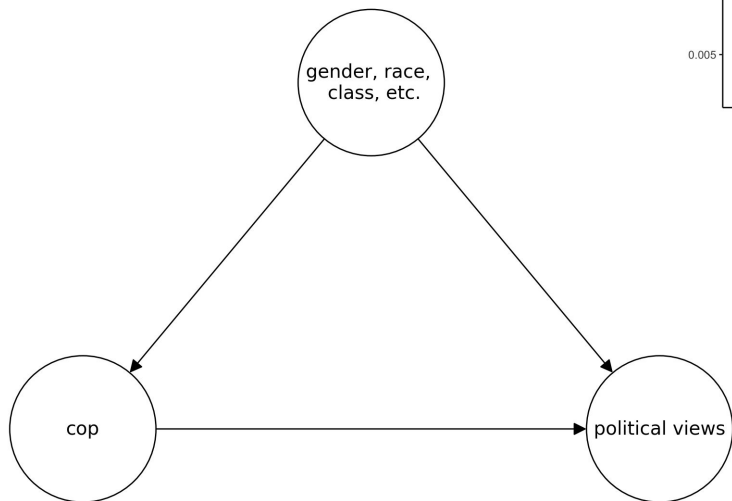
“a majority of Americans (64%) favor a ban on assault-style weapons” vs. “a similar share of police officers (67%) say they would oppose [it]”

- Mercer et al. (2017)

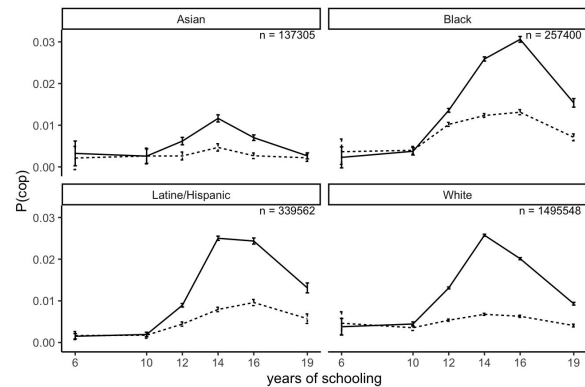
**What causes this difference?**

# THE PROBLEM

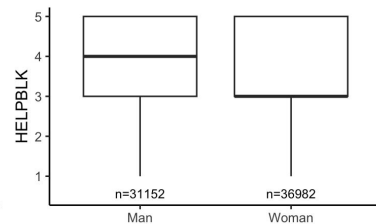
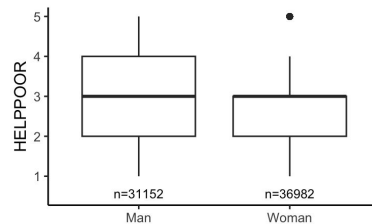
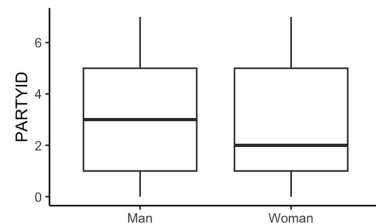
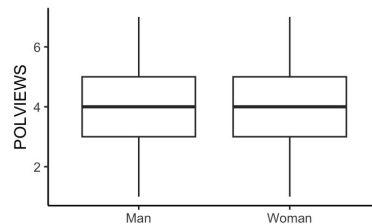
Not just anyone joins the police...



NOTE: race, age, and year of interview are held constant; n indicates cops at each level



NOTE: adult population, holding age constant



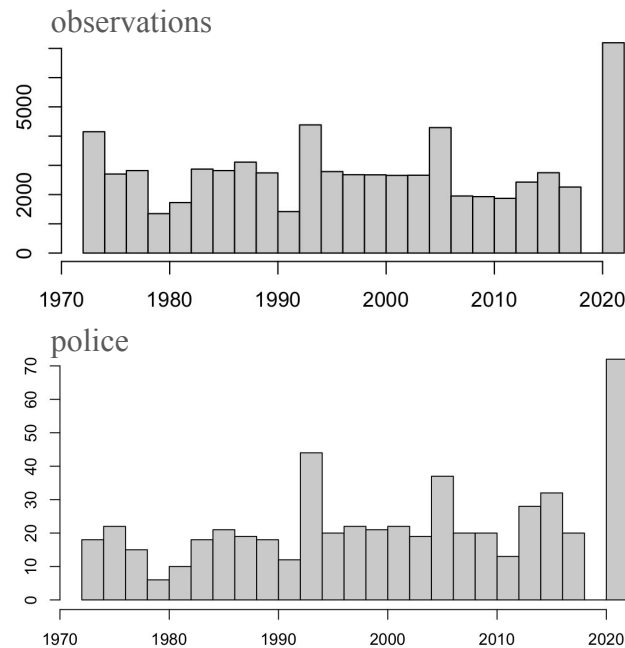
# THE DATA

- Ideal Data:
  - individual-level data
  - demographic variables, including detailed information on occupation
  - questions about political opinion
  - large  $n$ , because only a small subset of the population are police officers
  - panel data over many years, to get variation in the “treatment”

→ my data: General Social Survey (GSS)

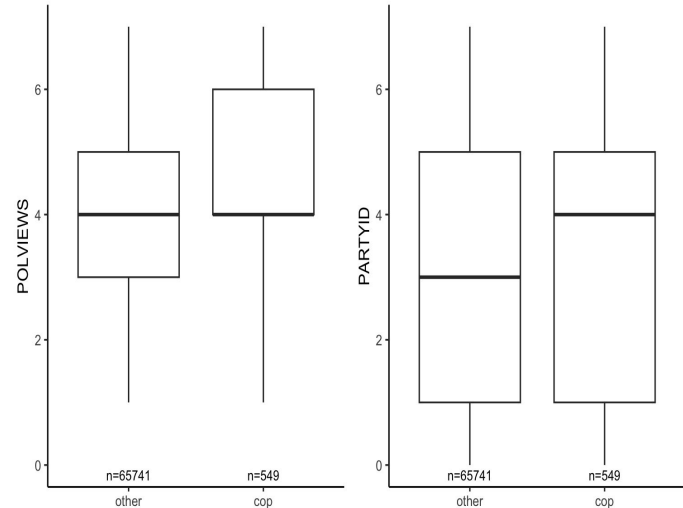
- NORC at UChicago
- 68,200 individuals, 1972 - 2022
- occupation codes, demographic, and opinion questions
- 549 police officers
  - incl. correctional officers

... over many years, but not a panel



# GENERAL SOCIAL SURVEY (GSS)

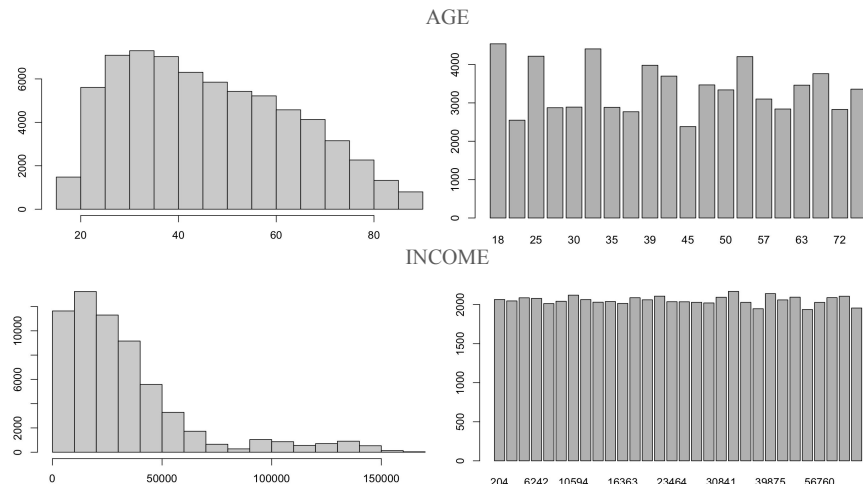
- 24 demographic controls
  - age, sex, race, household income, occupational prestige, father's occupational prestige, highest degree, mother's highest degree, father's highest degree, work status, employed/ self-employed, father's work status (when young), marital status, year-, and region of interview (proxy for residence), region of origin (at 16 yrs), which religion, practicing religion, rural/urban (9 levels), lived w/ family at 16, born in the US, parents born in the US, general happiness, and satisfaction w/ financial situation.
- opinion questions: Where would you place yourself on a scale--
  - POLVIEWS: ... from extremely liberal – point 1 – to extremely conservative – point 7?
  - PARTYID: ... from strong Democrat – point 0 – to strong Republican – point 6?



# THE METHOD

- The perfect comparison group: people who are exactly the same on all measurable characteristics
  - categorical controls : adding all main effects and all possible interactions
  - but for 24 variables, most of which are categorical à 2-9 categories... **too many dimensions!**
- Post-Double Selection (PDS) – Belloni et al. (2014); Angrist & Frandsen (2022)
  1. LASSO all controls on treatment –  $d \sim Z$
  2. LASSO all controls on outcome –  $y \sim Z$
  3. Keep controls selected in either step for final model –  $Y \sim D + Z$

cv.glmnet() 10-fold CV to select  $\lambda$   
 $\alpha = 1$
- binned continuous variables, then dummy encoding
  - year of interview, HH income, age, father's - / occupational prestige score (0-100) – 14 - 30 equal-size bins  
→ 170 main effects
- all possible interactions, removing empty columns  
→ 13,291 variables
- removing near-zero variance *interactions*
  - freqCut = 200, uniqueCut = 0.01  
→ 5,458 variables.



# RESULTS

$$\text{POLVIEWS/ PARTYID} \sim \text{cop} + \text{income} + \text{age} + \text{prestg} + \text{prestg\_pop} + Z_i$$

-  $Z_i$  included 901 controls for POLVIEWS and xxx for PARTYID

OLM

cumulative logit model

NULL

<i>Predictors</i>	POLVIEWS				PARTYID			
	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>
cop	0.30	0.06	0.18 – 0.42	<b>&lt;0.001</b>	0.39	0.08	0.23 – 0.56	<b>&lt;0.001</b>
Observations	59470				67815			
R <sup>2</sup> / R <sup>2</sup> adjusted	0.000 / 0.000				0.000 / 0.000			

PDS

<i>Predictors</i>	POLVIEWS				PARTYID			
	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>
cop	0.36	0.08	0.21 – 0.52	<b>&lt;0.001</b>	0.39	0.11	0.17 – 0.61	<b>0.001</b>
Observations	34160				35417			
R <sup>2</sup> / R <sup>2</sup> adjusted	0.193 / 0.161				0.244 / 0.214			

$$\text{logit}[P(Y \leq k)] = \beta_k + \text{cop} + \dots Z_i,$$

where  $k$  are the levels of the outcome.

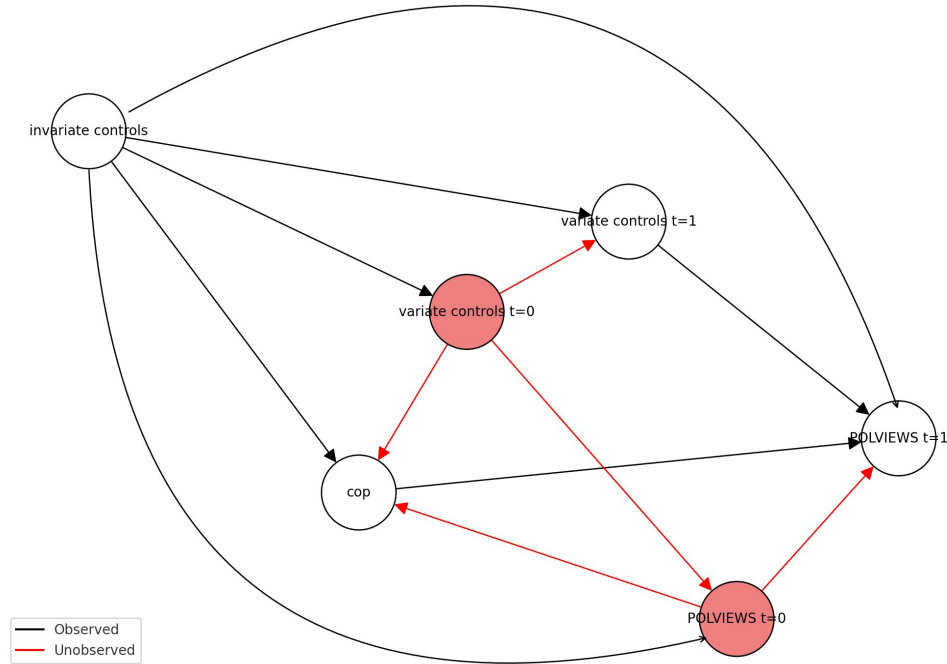
<i>Predictors</i>	POLVIEWS				PARTYID			
	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>std. Error</i>	<i>CI</i>	<i>p</i>
cop	-0.53	0.11	-0.75 – -0.32	<b>&lt;0.001</b>	-0.38	0.11	-0.59 – -0.16	<b>0.001</b>
Observations	34160				35417			

The odds of being one step more liberal are  $1 - e^{-0.53} = 0.41$ , i.e. 41% smaller for police than for non-police.

The odds of being one step more Democrat are  $1 - e^{-0.38} = 0.32$ , i.e. 32% smaller for police than for non-police.

# CONCLUSION

- Police are more conservative and more republican than non-police.
- The differences are not driven by *any* measurable confounding variables.
- Any remaining difference is only explainable by either, or both, of:
  1. ideological selection into police
  2. socialization effect upon joining the police.

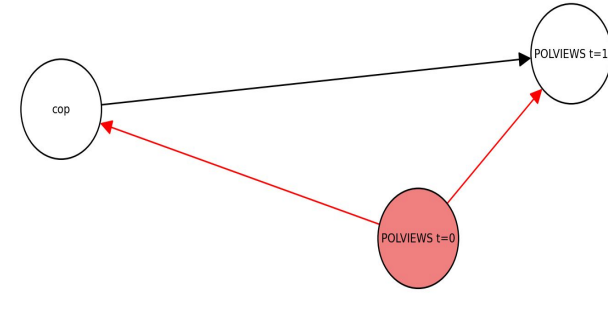




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- to understand the relative importance of causality here, we would actually need panel data...



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- Mercer, Rich Morin, Kim Parker, Renee Stepler and Andrew. 2017. “6. Police Views, Public Views.” Pew Research Center. Retrieved September 30, 2024 ([www.pewresearch.org/social-trends/2017/01/11/police-views-public-views/](http://www.pewresearch.org/social-trends/2017/01/11/police-views-public-views/)).
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