

# Empirical Research on Economic Inequality

## Experiments to test for discrimination in hiring

Maximilian Kasy

Harvard University, fall 2015

## Inequality between groups

- ▶ We observe large economic inequalities along dimensions such as race and gender.
- ▶ Why?
- ▶ Many channels through which they might be created!

## Possible channels

### Differences in

1. early childhood influences
2. neighborhoods of growing up
3. access to / quality of  
primary, middle, and high school education
4. chance of being hired when applying for a job
5. wages conditional on being hired
6. chance of being promoted or fired in a given job
7. treatment by customers or clients
8. treatment by police and courts
9. ...

## 4. Chance of being hired when applying for a job

Decomposes further into

- a. chance of being invited to an interview
- b. chance of being hired given an interview

## a. Chance of being invited to an interview

*Bertrand, M. and Mullainathan, S. (2004). Are Emily and Greg More Employable Than Lakisha and Jamal? A Field Experiment on Labor Market Discrimination. American Economic Review, 94(4):991–1013.*

- ▶ Chance might depend on
  1. the (perceived) race and gender of an applicant,
  2. neighborhood of residence,
  3. the high school attended, ...
- ▶ Bertrand and Mullainathan (2004):  
What is the causal effect of perceived race on the chance of being invited to an interview, for *otherwise identical* CVs?

## What is a causal effect?

- ▶ Potential outcome framework: answer to “what if” questions
- ▶ Two “treatments”:  $D = 0$  or  $D = 1$
- ▶ e.g. “black name” vs. “white name” on the CV
- ▶  $Y_i$ : CV  $i$ 's outcome  
e.g. being invited for an interview
- ▶ Potential outcome  $Y_i^0$ :  
what if CV  $i$  had a “black name” (treatment 0)
- ▶ Potential outcome  $Y_i^1$ :  
what if CV  $i$  had a “white name” (treatment 1)

## Questions for you

Does the “what if” question make sense?

After all, we can never observe what would have happened!

- ▶ Causal effect / treatment effect for CV  $i$  :

$$Y_i^1 - Y_i^0.$$

- ▶ Average causal effect / average treatment effect:

$$ATE = E[Y^1 - Y^0],$$

- ▶ Expectation averages over the population of interest.



## The fundamental problem of causal inference

- ▶ **We never observe both  $Y^0$  and  $Y^1$  at the same time.**
- ▶ One of the potential outcomes is always missing from the data.
- ▶ Treatment  $D$  determines which of the two we observe.
- ▶ Formally:

$$Y = D \cdot Y^1 + (1 - D) \cdot Y^0.$$

## Selection problem

- ▶ Distribution of  $Y^1$  among those with  $D = 1$  need not be the same as the distribution of  $Y^1$  among everyone.
- ▶ In particular

$$E[Y|D = 1] = E[Y^1|D = 1] \neq E[Y^1]$$

$$E[Y|D = 0] = E[Y^0|D = 0] \neq E[Y^0]$$

$$E[Y|D = 1] - E[Y|D = 0] \neq E[Y^1 - Y^0] = ATE.$$

- ▶ e.g., for real job applicants, race correlates with neighborhood, school, etc. ...

## Randomization

- ▶ No selection  $\Leftrightarrow D$  is random

$$(Y^0, Y^1) \perp D.$$

- ▶ In this case,

$$E[Y|D = 1] = E[Y^1|D = 1] = E[Y^1]$$

$$E[Y|D = 0] = E[Y^0|D = 0] = E[Y^0]$$

$$E[Y|D = 1] - E[Y|D = 0] = E[Y^1 - Y^0] = ATE.$$

- ▶ Can ensure this by actually randomly assigning  $D$ .
- ▶ Independence  $\Rightarrow$  comparing treatment and control actually compares “apples with apples.”
- ▶ This gives **empirical content** to the “metaphysical” notion of **potential outcomes!**

## Estimation

- ▶ Easy for randomized experiments
- ▶ Recall

$$ATE = E[Y_1 - Y_0] = E[Y|D = 1] - E[Y|D = 0].$$

- ▶ Estimator:

$$\hat{\alpha} = \bar{Y}_1 - \bar{Y}_0,$$

where

$$\bar{Y}_1 = \frac{\sum Y_i \cdot D_i}{\sum D_i} = \frac{1}{N_1} \sum_{D_i=1} Y_i$$

$$\bar{Y}_0 = \frac{\sum Y_i \cdot (1 - D_i)}{\sum (1 - D_i)} = \frac{1}{N_0} \sum_{D_i=0} Y_i.$$

## Questions for you

Show that

$$E[\hat{\alpha}] = ATE$$

if  $(Y^0, Y^1) \perp D$ .

## Inference

- ▶ Range of likely values for ATE?
- ▶  $t$ -statistic:

$$t = \frac{\hat{\alpha} - \alpha_{ATE}}{\hat{\sigma}_{\alpha}}$$

where

$$\hat{\sigma}_{\alpha} = \sqrt{\frac{\hat{\sigma}_1^2}{N_1} + \frac{\hat{\sigma}_0^2}{N_0}}$$

and

$$\hat{\sigma}_1^2 = \frac{1}{N_1 - 1} \sum_{D_i=1} (Y_i - \bar{Y}_1)^2.$$

- ▶  $\hat{\sigma}_0^2$  is analogously defined.

## Confidence interval

- ▶  $t$ -statistic is approximately standard normal distributed (for samples of a reasonable size),

$$t \sim^{approx} N(0, 1).$$

- ▶ 95% confidence interval:

$$CI = [\hat{\alpha} - 1.96 \cdot \hat{\sigma}_{\alpha}, \hat{\alpha} + 1.96 \cdot \hat{\sigma}_{\alpha}].$$

## Questions for you

Show that

$$P(\alpha \in CI) \approx 0.95.$$

(Homework)

Note that  $\alpha$  is fixed, while  $CI$  is random!