IVII & RDDII

PMAP 8521: Program Evaluation for Public Service November 18, 2019

Fill out your reading report on iCollege!

PLAN FOR TODAY

Instruments

Treatment effects and compliance

Fuzzy RD

Synthetic data with R

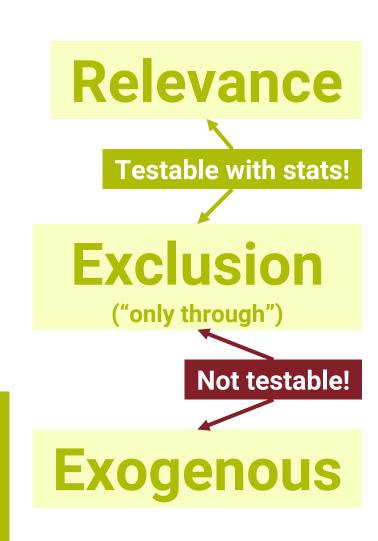
INSTRUMENTS

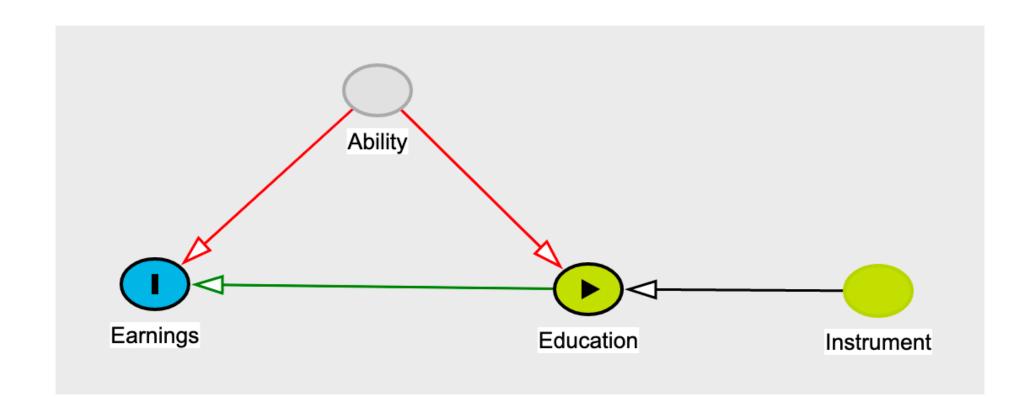
WHAT IS AN INSTRUMENT?

Something that is correlated with the policy variable

Something that does not directly cause the outcome

Something that is not correlated with the omitted variables





TREATMENT EFFECTS & COMPLIANCE

POTENTIAL OUTCOMES

$$\delta = (Y|P=1) - (Y|P=0)$$

 δ = Causal impact of program

P = Program

Y = Outcome

$$\delta = Y_1 - Y_0$$

Fundamental problem of causal inference

$$\delta_i = Y_i^1 - Y_i^0$$

Individual-level effects are impossible to observe

AVERAGE TREATMENT EFFECT

Difference between expected value when program is on vs. expected value when program is off

$$ATE = E(Y_1 - Y_0) = E(Y_1) - E(Y_0)$$

Can be found for a whole population, on average

$$\delta = (\bar{Y}|P=1) - (\bar{Y}|P=0)$$

Every individual has a treatment/causal effect

ATE = average of all unit-level causal effects

ATE = average effect for the whole population

VERSIONS OF CAUSAL EFFECTS

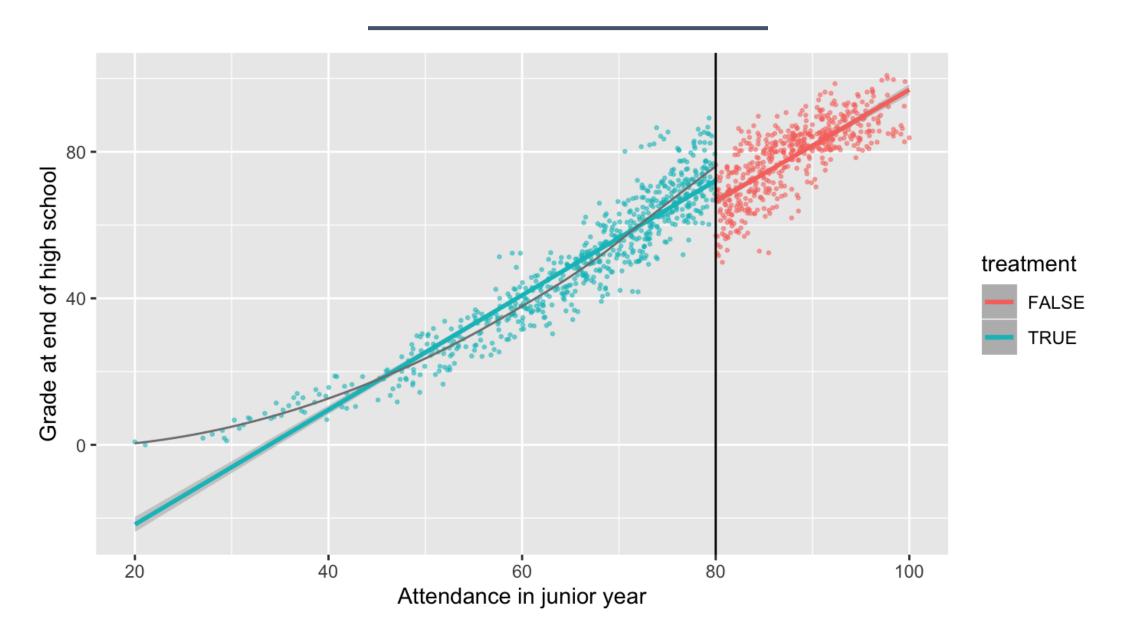
Average treatment on the treated

ATT / TOT

Conditional average treatment effect

CATE

LOCAL EFFECTS



LATE

Local average treatment effect (LATE) = weighted ATE

Narrower effect; only includes some of the population

Can't make population-level claims with LATE

(But that can be okay)

LATE

In RDD, LATE = people in the bandwidth

In RCTs, IVs, etc., LATE = compliers

COMPLIANCE

Compliers

Treatment follows assignment

Always takers

Gets treatment regardless of assignment

Never takers

Rejects treatment regardless of assignment

Defiers

Does opposite treatment from assignment









Always takers







N N

Never takers









Compliers

IGNORING DEFIERS

We can generally assume defiers don't exist

In drug trials this makes sense; can't get access to medicine without being in treatment

In development, it can make sense; in a bed net RCT, a defier assigned to treatment would have to tear down all existing bed nets out of spite

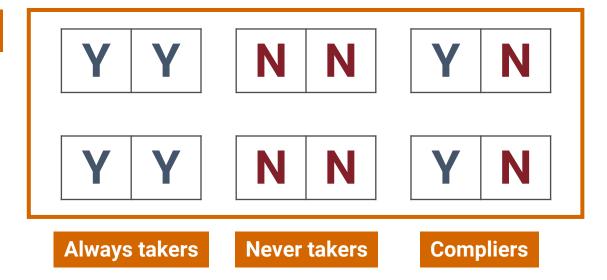
IGNORING DEFIERS

Monotonicity assumption

Assignment to treatment only has an effect in one direction

Assignment to treatment can only increase—not decrease—your actual chance of treatment





Assigned to treatment

Always takers & compliers





Never takers





Assigned to control





Always takers



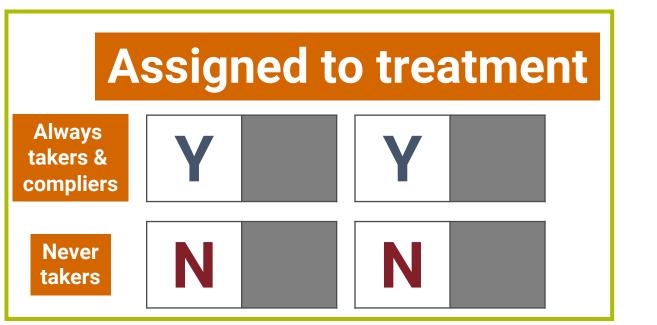


Never takers & compliers

MORE EFFECTS

Intent to treat (ITT)

Effect of assignment (not actual treatment!





MORE EFFECTS

Complier Average Causal Effect (CACE)

LATE for the compliers

Assigned to treatment













Assigned to control







N



Never takers & compliers

Assigned to treatment

Assigned to control

Always takers & compliers





Never takers





ITT =
$$\pi_{\text{compliers}} \times (T - C)_{\text{compliers}} +$$

$$\pi_{\text{always takers}} \times (T - C)_{\text{always takers}} +$$

$$\pi_{\text{never takers}} \times (T - C)_{\text{never takers}}$$

ITT =
$$\pi_{\rm C}$$
CACE + $\pi_{\rm A}$ ATACE + $\pi_{\rm N}$ NTACE

ITT =
$$\pi_{\rm C}$$
CACE + $\pi_{\rm A}$ ATACE + $\pi_{\rm N}$ NTACE

$$ITT = \pi_{C}CACE + \pi_{A}0 + \pi_{N}0$$

Exclusion restriction; treatment received is same regardless of assignment

$$ITT = \pi_{C}CACE$$

$$CACE = \frac{ITT}{\pi_C}$$

$$CACE = \frac{ITT}{\pi_C}$$

$$ITT = (\bar{y}|Treatment) - (\bar{y}|Control)$$

Assigned to treatment













Assigned to control





$$\pi_{\rm A} + \pi_{\rm C} = \%$$
 in treatment and yes $\pi_{\rm C} = \%$ in treatment and yes $-\pi_{\rm A}$

$$CACE = \frac{ITT}{\pi_C}$$

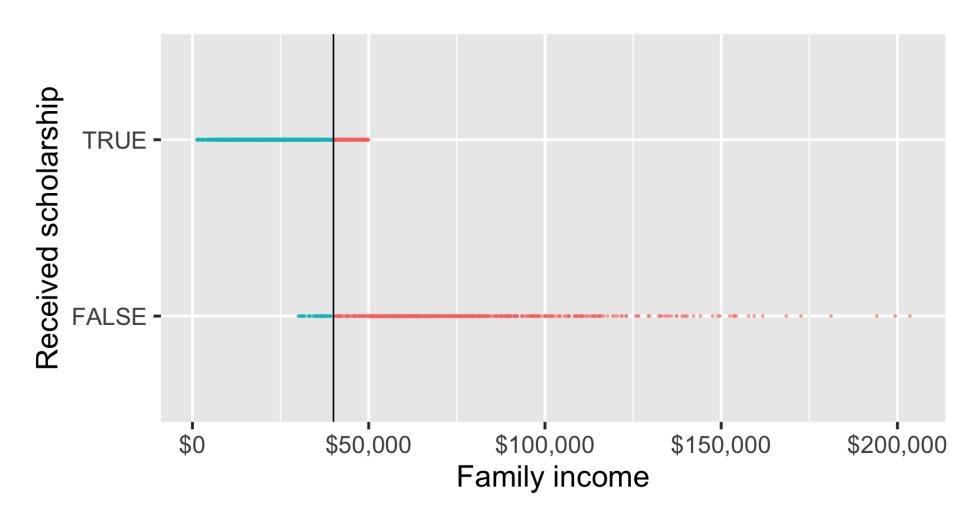
$$ITT = (\bar{y}|Treatment) - (\bar{y}|Control)$$

$$\pi_{\rm C} = \%$$
 in treatment and yes— $\%$ in control and yes

Example in R

FUZZY RD

oh no



Income < \$40,000: ● FALSE ● TRUE

Use an instrument to deal with noncompliance

Often actual participation in program works as instrument

SYNTHETIC DATA WITH R