

Topics in Causal Inference

STAT41530

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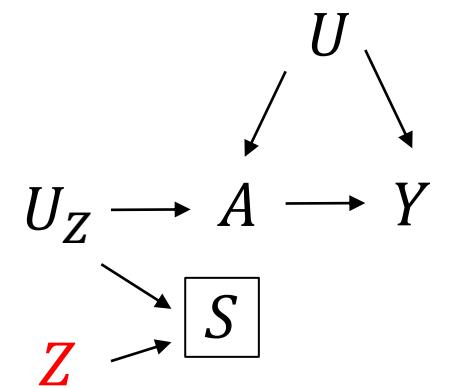
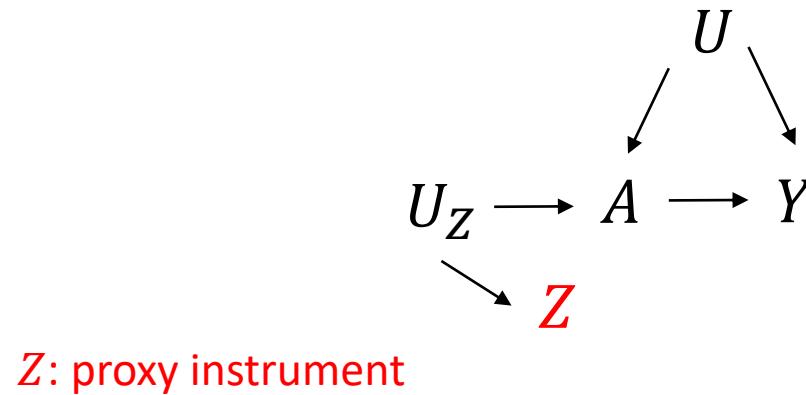
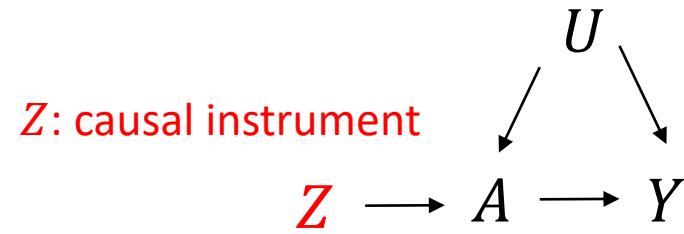
Lecture 7

Topic:
Instrumental Variables (IV)

- Three conditions for valid IV
- Identifiability conditions using IV
 - Homogeneity
 - Monotonicity (principal stratification)
- Weak IV

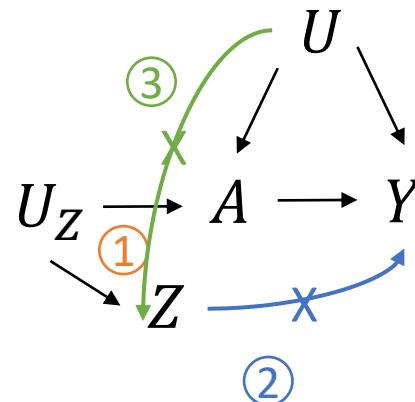
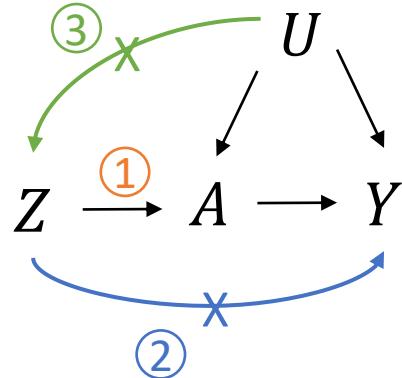
Instrumental variables (IV)

Identify causal effects without conditional exchangeability?



- Random assignment not available for treatment A , but for some other variable Z
- Non-compliance in randomized experiments

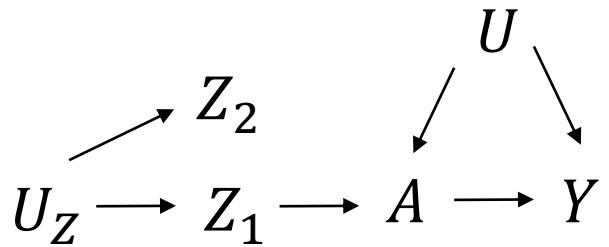
Three conditions for valid IV



- 1 Relevance: $Z \perp\!\!\!\perp A$, Z is associated with A
- 2 Exclusion restriction: Z does not affect Y except through potential effects on A .
Structural equation $Y = f(A, Z, U, E_Y) = f(A, U, E_Y)$
- 3 Random assignment: Z and Y do not have any common causes, $Z \perp\!\!\!\perp Y(a) \mid L$ for all a

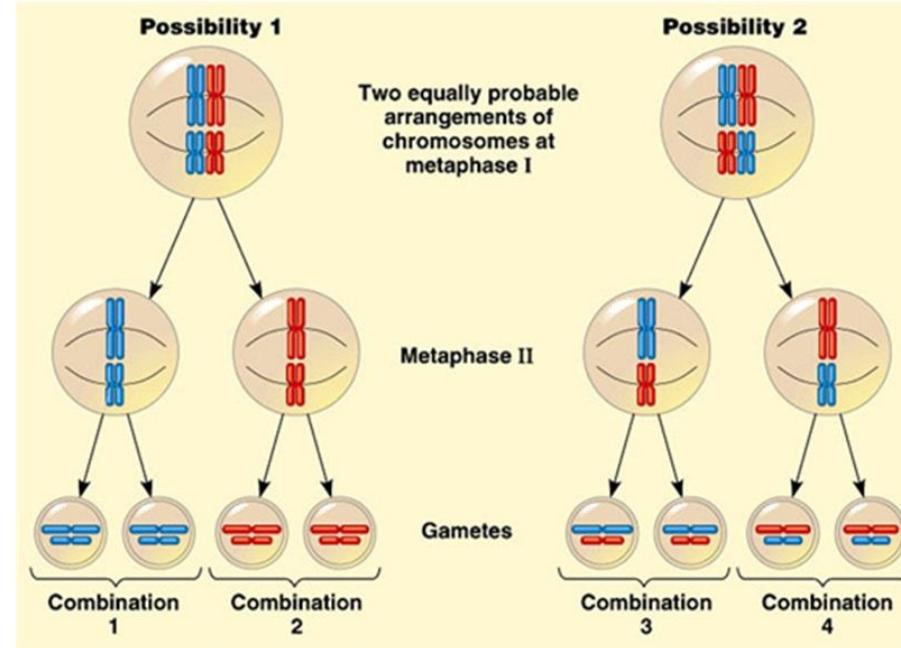
Examples of IV

- Genetic mutations (Mendelian Randomization)



- Access to treatment
 - Treatment assignment
 - Encouragement / Rewards
 - Cigarettes' price
 - Distance to the hospital

Mendel's law of Inheritance INDEPENDENT ASSORTMENT



<http://fig.cox.miami.edu/~cmallery/150/mitosis/c13x9independent-assortment.jpg>

Valid IV \Rightarrow Identification of causal effect of A ?

NO!

- With valid IV, we can only identify the upper/lower bounds of the causal effect of A on Y
(Robins 1989, Manski 1990)

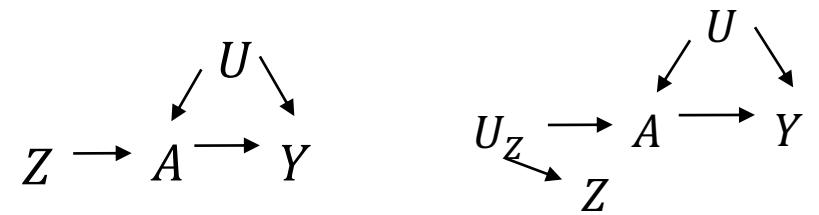
If Z, A, Y are binary ,then the range of $\tau = E[Y(1)] - E[Y(0)]$ has width

$$P(A = 1 | Z = 0) + P(A = 0 | Z = 1)$$

- We need either of the two further assumptions
 - Homogeneity
 - Monotonicity
- In this lecture, we focus on the case where A, Z are both binary
(Palmer 2011 reviewed more complicated scenarios)

The standard IV estimator

- The usual IV estimand



$$\tau_{IV} = \frac{E[Y | Z = 1] - E[Y | Z = 0]}{E[A | Z = 1] - E[A | Z = 0]} = \frac{\text{Cov}(Y, Z)}{\text{Cov}(A, Z)}$$

The IV estimand bypass the need to adjust for the confounder by inflating the intention-to-treat effect

- The standard IV estimator

$$\hat{\tau} = \frac{\hat{E}[Y | Z = 1] - \hat{E}[Y | Z = 0]}{\hat{E}[A | Z = 1] - \hat{E}[A | Z = 0]} = \frac{\frac{1}{n_1} \sum Y_i 1_{Z_i=1} - \frac{1}{n_0} \sum Y_i 1_{Z_i=0}}{\frac{1}{n_1} \sum A_i 1_{Z_i=1} - \frac{1}{n_0} \sum A_i 1_{Z_i=0}}$$

- Two-step least square

- Step 1: $E[A | Z] = \alpha_0 + \alpha_1 Z$ regress A on Z , get $\hat{E}[A | Z]$

- Step 2: $E[Y | Z] = \gamma_0 + \gamma_1 Z = \frac{\gamma_1}{\alpha_1}(\alpha_0 + \alpha_1 Z) + \gamma_0 - \frac{\alpha_0 \gamma_1}{\alpha_1} = \tau_{IV} E[A | Z] + \gamma_0 - \frac{\alpha_0 \gamma_1}{\alpha_1}$
regress Y on $\hat{E}[A | Z]$

When does the IV estimand have causal interpretation?

Homogeneity assumption

- Robins mentioned 4 possible homogeneity conditions (section 16.3)
- The simplest (most stringent) condition: **constant treatment effect of A**

$$Y(a) = f(a, U, E_Y) = \tau a + \tilde{f}(U, E_Y)$$

- This means that for any individual: $Y_i(1) - Y_i(0) = \tau$

Such homogeneity assumption is impossible for binary outcome (unless $\tau = 0$)

$$\text{As } Y = \tau A + \tilde{f}(U, E_Y), \quad Z \perp \tilde{f}(U, E_Y)$$

$$\text{Then } \mathbb{E}[Y | Z] = \mathbb{E}[\tau A + \tilde{f}(U, E_Y) | Z] = \tau \mathbb{E}[A | Z] + \mathbb{E}[\tilde{f}(U, E_Y)]$$

$$\tau_{IV} = \frac{\mathbb{E}[Y | Z = 1] - \mathbb{E}[Y | Z = 0]}{\mathbb{E}[A | Z = 1] - \mathbb{E}[A | Z = 0]} = \tau$$

- A milder homogeneity condition for identifying $\tau = E[Y(1)] - E[Y(0)]$:

The average causal effect is the same at different level of U

$$E[Y(1) | U] - E[Y(0) | U] \equiv \tau$$

Homogeneity assumption example

- Non-constant treatment effect

Structural equation (let $E_Y = (E_0, E_1)$, $E(E_Y) = 0$)

$$Y = \beta_0 + \tau A + U + AE_1 + (1 - A)E_0$$

corresponding potential outcomes:

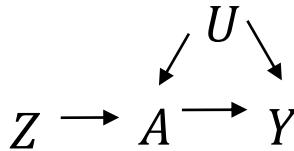
$$Y(0) = \beta_0 + U + E_0, \quad Y(1) = \beta_0 + \tau + U + E_1$$

treatment effect: $Y(1) - Y(0) = \tau + E_1 - E_0$

- Identification of average causal effect: $\tau_{IV} = \tau$ as $E[Y | Z] = \tau E[A | Z] + E[U]$
- Notice that the structural equation $Y = \beta_0 + \tau A + U + E_Y$ is different from the structural equation $Y = \beta_0 + \tau A + U + AE_1 + (1 - A)E_0$
 - If $E_0 \sim E_1$, then the two structural equations are not distinguishable from the observed data
 - The noise E_Y can have one or multiple dimensions

Principal stratification

- Only works for binary causal instrument



- Define potential outcomes of Z : $A(0), A(1), Y(Z = 0), Y(Z = 1)$
- Partition individuals into 4 groups (principal stratification)

		Groups (G)
$A(1) = 1$	$A(0) = 1$	Always takers (a)
$A(1) = 1$	$A(0) = 0$	Complier (c)
$A(1) = 0$	$A(0) = 1$	Never taker (n)
$A(1) = 0$	$A(0) = 0$	Defier (d)

- Define intention to treatment $ITT_{Y,g} = E[Y(Z = 1) - Y(Z = 0)|G = g]$ and the proportion of each group $\pi_g = P(G = g)$
- $ITT_{Y,a} = E[Y(1) - Y(1)|G = a] = 0 = E[Y(0) - Y(0)|G = n] = ITT_{Y,0}$
- Then the intention to treatment effect is

$$E[Y(Z = 1)] - E[Y(Z = 0)] = \pi_c ITT_{Y,c} + \pi_d ITT_{Y,d}$$

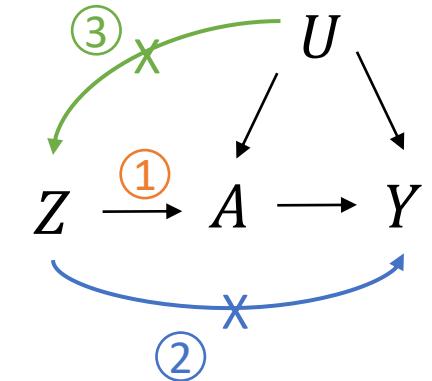
Monotonicity assumption

- Relevance $\Rightarrow \pi_c \neq 0$
- Exclusion restriction $\Rightarrow ITT_{Y,a} = ITT_{Y,n} = 0$
- Random assignment $\Rightarrow E[Y(Z = 1)] - E[Y(Z = 0)] = E[Y|Z = 1] - E[Y|Z = 0]$
- **Monotonicity (no defiers):** $\pi_d = 0$
- Average causal effect of A on Y for the compliers (CACE: complier average causal effect):

$$E[Y(1) - Y(0)|G = c] = E[Y(Z = 1) - Y(Z = 0)|G = c] = ITT_{Y,c}$$
- As $\pi_d = 0$,

$$ITT_{Y,c} = \frac{E[Y(Z = 1)] - E[Y(Z = 0)]}{\pi_c} = \frac{E[Y|Z = 1] - E[Y|Z = 0]}{E[A|Z = 1] - E[A|Z = 0]} = \tau_{IV}$$

$$\begin{aligned} \mathbb{E}[A | Z = 1] - \mathbb{E}[A | Z = 0] &= \mathbb{E}[A(1)] - \mathbb{E}[A(0)] \\ &= \pi_a \cdot 0 + \pi_c \cdot 1 + \pi_n \cdot 0 + \pi_d \cdot (-1) = \pi_c \end{aligned}$$



Weak instrument

- π_c is small ($\text{Cov}(A, Z)$ is small): sensitive to violations of valid IV
- $\hat{E}[A | Z]$ is inaccurate as an estimator of $E[A | Z]$
two much uncertainty in the first-step \Rightarrow weak IV bias in the the two-step least square
- Rule of thumb: an instrument is considered weak if the F statistics for the $A - Z$ association is **less than 10**