

Causal assumptions

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- (1) SUTVA (↳ stable unit treatment value assumption)
- (2) Consistency
- (3) Ignorability
- (4) Positivity

} required for identifiability of causal effects
} untestable assumptions

- these assumptions refer to the observed data: $\{Y, A, X\}$

outcome ← treatment ↓ pre-treatment covariates X

1) SUTVA: (involves 2 assumptions)

1. No interference: Units do not interfere w/ each other [i.e. the treatment assignment of one unit doesn't affect the treatment effectiveness / outcome / treatment assignment of another unit]



another word: no spillover or contagion b/w units

2. One version of treatment: There is only one variable we can hypothetically intervene on. The treatment is very well defined.

⇒ SUTVA "decouples" units in the observational dataset

2) Consistency:

- Potential outcome under treatment $A=a$, Y^a , is equal to the observed outcome Y , if the actual treatment received is $A=a$

↳ in other words: $Y = Y^a$ if $A=a$, for all a ($\forall a \in A$)

3) Ignorability: (sometimes: "no unmeasured confounders" assumption)

- given pre-treatment covariates X , treatment assignment is independent of potential outcomes

↳ $(Y^0, Y^1) \perp\!\!\!\perp A | X$ "treatment assignment ignorable given confounders X " means: within levels of X , treatment assignment is independent from the potential outcomes.

among people/units with covariates X , this allows us to think of treatment

A as being randomly assigned among that subpopulation

↳ example: $Y^0, Y^1 \perp\!\!\!\perp A$ (because $X \rightarrow A \rightarrow Y$ & $X \rightarrow Y$)

- X ... a single variable (age) with values {young, old}

- fact 1: old people more likely to get treatment $A=1$

- fact 2: old people also more likely to have the outcome (hip fracture) Y , regardless treatment A

⇒ THEN: Y^0 and Y^1 aren't independent from A (marginally)

⇒ However, within levels of X , treatment might be randomly assigned!

4) Positivity assumption:

- every unit had some chance of receiving any treatment a ($\forall a, \forall x$)

↳ $P(A=a | X=x) > 0$ ($\forall a, \forall x$)

• means that treatment assignment is not deterministic (e.g. as a fct. of X)

↳ if for some values/levels of X , treatment was deterministic, then we would have no observed values of Y for one of the treatment groups for those values/levels of X

⇒ in short: we need variability in treatment assignment for successful treatment effect identifiability!

Applying these assumptions to causal effect estimation:

$$Y = Y^a \text{ if } A=a$$

$$(1) E[Y | A=a, X=x] = E[Y^a | A=a, X=x] \text{ by consistency assumption in (2).}$$

$$= E[Y^a | X=x] \text{ by ignorability assumption in (3)}$$

$$Y^0, Y^1 \perp\!\!\!\perp A | X$$

$$\Rightarrow E[Y | A=a, X=x] = E[Y^a | X=x]$$

↓ marginalize over distribution of X

$$E[Y^a]$$

Quiz answers:

(1) "No unmeasured confounders" \equiv Ignorability

(2) "No interference between units" \equiv SUTVA