

Causal Inference Chap 7: Confounding

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July 29, 2021

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The structure of confounding

Confounding: form of lack of exchangeability between the treated and the untreated.

→ Presence of confounding holds "association is not causation"

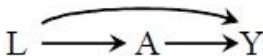


Figure 7.1

- Figure shows treatment A outcome Y and common cause L
- The path $A \leftarrow L \rightarrow Y$ is an example of a *backdoor path*.
- Presence of the common cause L creates a confounding for the effect of A on Y .
- Hence, associational risk ratio is not equal to causal risk ratio

$$\frac{Pr[Y=1|A=1]}{Pr[Y=1|A=0]} \neq \frac{Pr[Y^{a=1}=1]}{Pr[Y^{a=0}=1]}$$

The structure of confounding

Examples of confounding in observational research

- Occupational factors: The effect of working as a firefighter A on the risk of death Y will be confounded if "being physically fit" L is a cause of both being a firefighter and having a lower mortality risk.
- Clinical decisions: The effect of drug A (aspirin) on the risk of disease Y (stroke) will be confounded if the drug is more likely to be prescribed to individuals with certain condition L (heart disease). And both L and Y are caused by atherosclerosis U , an unmeasured variable.(*confounding by indication*)

→ All the cases are because of the bias with same structure.
(common cause L results in open backdoor path between A and Y)

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① The structure of confounding

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Confounding and exchangeability

Question: If confounding is likely, can we determine whether there exists a set of measured covariates L for which conditional exchangeability holds?

→ thinking in terms of conditional exchangeability $Y^a \perp A|L$ is often not intuitive

- 1 the backdoor criterion applied to the causal DAG
- 2 the transformation of the causal DAG into a SWIG

Confounding and the backdoor criterion

Backdoor criterion

- ① A set of covariates L satisfies the *backdoor criterion* if all backdoor paths between A and Y are blocked by conditioning on L
- ② L contains no variables that are descendants of treatment A

→ conditional exchangeability $Y^a \perp A | L$ holds iff L satisfies the backdoor criterion

Confounding and the backdoor criterion

Relation between backdoor criterion and confounding

- ① If no common causes of treatment and outcome, the backdoor criterion is satisfied. This describes a marginally randomized experiment in which confounding is not expected.
- ② Common causes of treatment and outcome but a subset L of measured non-descendants of A suffices to block all backdoor paths. This describes a conditionally randomized experiment in which the probability of receiving treatment is the same for all individuals with the same value of L .

Confounding and the backdoor criterion

The application of the backdoor criterion to determine whether the causal effect of A on Y is identifiable and which variables are required to ensure conditional exchangeability.



Figure 7.1

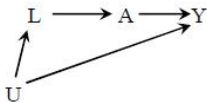


Figure 7.2

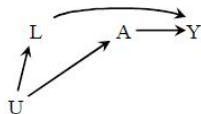


Figure 7.3

→ Backdoors can be blocked by conditioning on L or U .

Confounding and the backdoor criterion

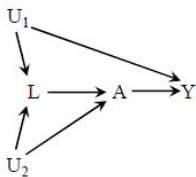


Figure 7.5

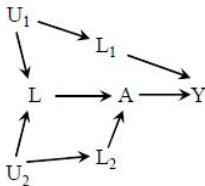


Figure 7.6

- In figure 7.5, the bias is intractable since conditioning on L would block the backdoor path ($A \leftarrow L \leftarrow U_1 \rightarrow Y$) but opens another backdoor path on which L is a collider ($A \leftarrow U_2 \rightarrow L \leftarrow U_1 \rightarrow Y$)
- The solution is to measure either (i) a variable L_1 between U_1 and Y (conditional exchangeability given L_1) or (ii) variable L_2 between U_2 and A (conditional exchangeability given both L and L_2) as in figure 7.6

Single-world intervention graphs

SWIG unifies the counterfactual and graphical approaches by explicitly including the counterfactual variables on the graph
→ SWIG depicts the variables and causal relations in a hypothetical world in which all individuals received treatment level a .

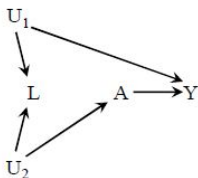


Figure 7.4

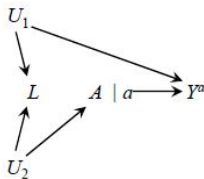


Figure 7.10

- Marginal exchangeability $Y^a \perp A$ holds because all paths between Y^a and A are blocked without conditioning on L
- Conditional exchangeability $Y^a \perp A | L$ not holds because the path is open when the collider L is conditioned on