

Causal Graphs*

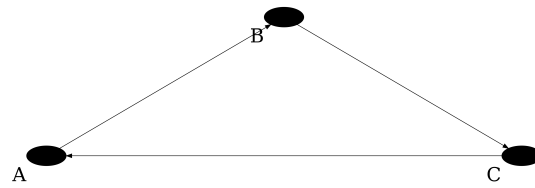
Definitions, patterns, and strategies

Definitions

- A **node** represents a random variable labeled by letter. Observed random variables are marked by solid circle ● and unobserved - by hollow circle ○.
- An **edge** shows dependence between joining variables.
- **Adjacent variables** are connected by an edge.
- **Adjacent edges** meet at a variable.
- A **directed edge** represents the cause by a single-headed arrow.
- A **parent/child** is the starting(tail)/ending(head) variable. Therefore, a directed edge represents a direct effect of a parent on a child.
- A **root** is a variable that has no parent. In other words, it is an exogenous variable determined only by forces outside of the graph.
- A **sink** is a variable with no children.
- A **path** is a sequence of adjacent edges.
- A **directed path** is a path traced out entirely along arrows tail-to-head. If there is a directed path from A to B, A is an **ancestor** of B; B is a **descendant** of A.
- A **directed acyclic graph (DAG)** is a graph with only arrows for edges and no feedback loops (i.e. no variable is its own ancestor or its own descendant):

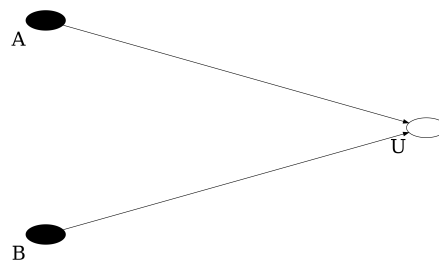
*We are indebted to Liudmila Kiseleva for the design of the handout.

Figure 1: Directed acyclical graph



- **Mutual dependence** of two variables on one or more common causes is shown by a curved and dashed bidirected edge:

Figure 2: Mutual dependence



Patterns

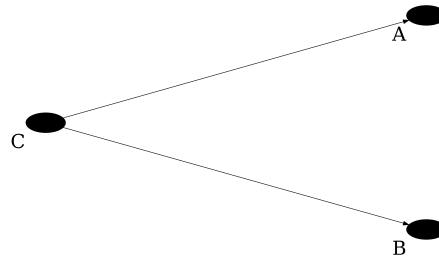
- **Chain of mediation** is a relationship when A affects B through A's causal effect on C and C's causal effect on B.

Figure 3: Mediation



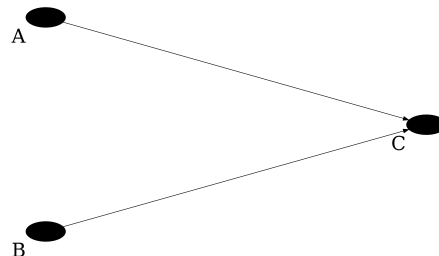
- **Mutual dependence** is a relationship when A and B are both caused by C.
- A **confounding variable** is a variable that affects both the dependent and independent variable.

Figure 4: Mutual dependence



- **Mutual causation** is a relationship when A and B are both causes of C.
- A **collider** is a variable that has two arrows running into it.

Figure 5: Mutual causation



- A **back-door path** is a path between any causally ordered sequence of two variables that include a directed edge that points to the first variable.
- **Conditioning** as a modeling strategy means transforming one graph into a simpler set of component graphs where fewer causes are represented.

Strategies

A **back-door criterion** is a set of conditions used to determine whether or not conditioning on a given set of observed variable will identify the causal effect. The causal

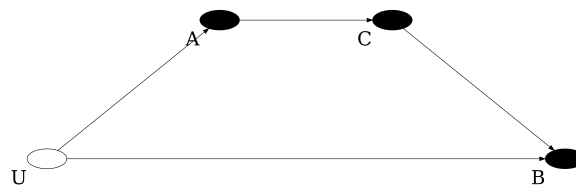
effect is identified by conditioning on a set of variables Z if and only if all back-door paths between the causal variable and the outcome variable are blocked after conditioning on Z . All back-door paths are blocked by Z if and only if each back-door path:

- contains a chain of mediation $A \rightarrow C \rightarrow B$ where the middle variable C is in Z , or
- contains a fork of mutual dependence $A \leftarrow C \rightarrow B$, where the middle variable C is in Z , or
- contains an inverted fork of mutual causation $A \rightarrow C \leftarrow B$, where the middle variable C and all of C 's descendants are not in Z .

A **front-door criterion** is an empirical strategy used to identify the causal relationship flowing from A to B if one can find a mechanism C which:

- lies on the causal path between A and B , and
- it is the only such mechanism, and
- it is not affected by the unobserved confounder U .

Figure 6: Mutual causation



You can find more on front-door criterion application in the Bellemare & Bloem (2020) paper.

References

Bellemare, M., & Bloem, J. (2020). The paper of how: estimating treatment effects using the front-door criterion.

Morgan, S. L., & Winship, C. (2014). *Counterfactuals and causal inference*. Cambridge, England: Cambridge University Press.

Pearl, J. (2009). *Causality*. Cambridge, England: Cambridge University Press.