

Section 5

Observational Studies 2

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GOV 2003

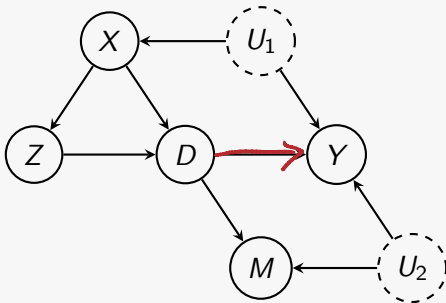
Oct 14, 2021

Overview

- Logistics:
 - **Pset 5 released!** Due at 11:59 pm (ET) on Oct 20
- Today's topics:
 1. Directed Acyclic Graphs

Motivation

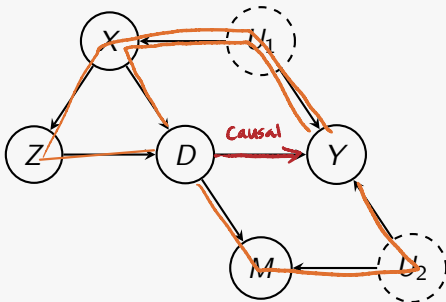
- Assume that you've come out with a DAG based on your expertise.



- Suppose you want to identify a causal effect of D on Y .

Motivation

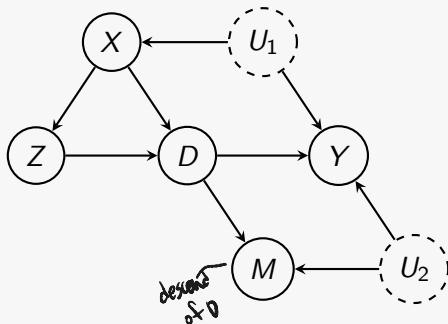
- Assume that you've come out with a DAG based on your expertise.



- Suppose you want to identify a causal effect of D on Y .
 - In a nutshell, what you might want to do is to block all the paths that yields statistical associations between D and Y .

Motivation

- Assume that you've come out with a DAG based on your expertise.



- Suppose you want to identify a causal effect of D on Y .
 - In a nutshell, what you might want to do is to block all the paths that yields statistical associations between D and Y .
 - Thus, you want to find a set of nodes S such that
 - once we condition on S , **no unmeasured confounding** holds and
 - any descend of D is not in $S \leadsto$ **no post-treatment bias**.
- \leadsto Use backdoor criterion!

DAG

- Things we have to know to check the backdoor criterion:

DAG

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 - Three common structures:
 - confounder (fork): $A \leftarrow C \rightarrow B$
 - collider (inverted fork): $A \rightarrow C \leftarrow B$
 - mediator (chain): $A \rightarrow C \rightarrow B$

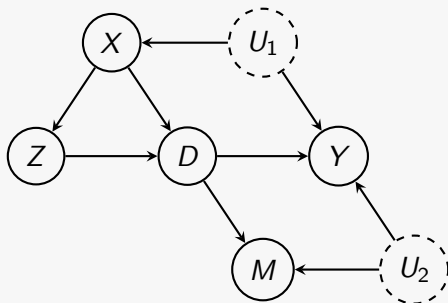
DAG

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 - Three common structures:
 - confounder (fork): $A \leftarrow C \rightarrow B$
 - collider (inverted fork): $A \rightarrow C \leftarrow B$
 - mediator (chain): $A \rightarrow C \rightarrow B$
 - How to block a path between A and B
 - If $A \leftarrow C \rightarrow B$: condition on C .
 - If $A \rightarrow C \leftarrow B$: **do not** condition on C .
 - If $A \rightarrow C \rightarrow B$: condition on C .

DAG

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 - If $A \rightarrow C \leftarrow B$: **do not** condition on C .
 - If $A \rightarrow C \rightarrow B$: condition on C .
 - **D-separation**: $A \perp\!\!\!\perp B \mid C$
 1. Find all paths between A and B .
 2. Check if each path is **blocked**.
 3. If all paths are blocked, then A is **d-separated** from B by C

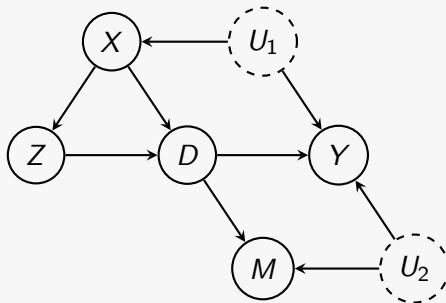
Backdoor criterion example



1. List all of the **backdoor** paths between D and Y .

- $D \leftarrow X \leftarrow U_1 \rightarrow Y$
- $D \leftarrow Z \leftarrow X \leftarrow U_1 \rightarrow Y$

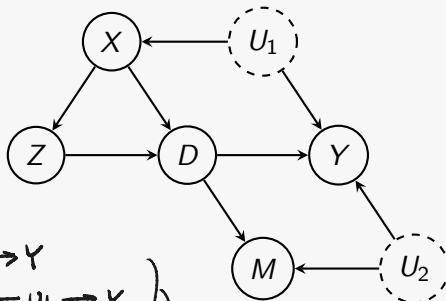
Backdoor criterion example



1. List all of the **backdoor** paths between D and Y .
2. List all the possible set of nodes \mathbf{S} that you can condition on.

\emptyset $\{X\}$ $\{Z\}$ $\{M\}$
 $\{X, Z\}$ $\{X, M\}$ $\{Z, M\}$
 $\{X, Z, M\}$

Backdoor criterion example

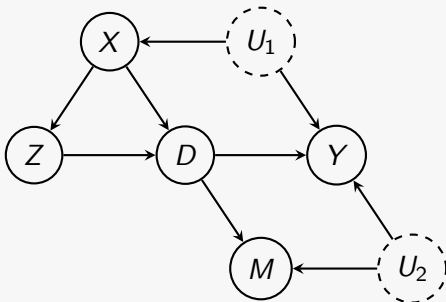


- $D \leftarrow \boxed{X} \leftarrow U_1 \rightarrow Y$
- $D \leftarrow \boxed{Z} \leftarrow \boxed{X} \leftarrow U_1 \rightarrow Y$ }

1. List all of the **backdoor** paths between D and Y .
2. List all the possible set of nodes **S** that you can condition on.
3. List all the **S** such that **blocks** all the backdoor paths.

\emptyset $\{X\}$ $\{Z\}$ $\{M\}$
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Backdoor criterion example



1. List all of the **backdoor** paths between D and Y .
2. List all the possible set of nodes **S** that you can condition on.
3. List all the **S** such that **blocks** all the backdoor paths.
4. Among those **S**, drop the sets which include a descend of D .

\mathcal{M}

R package: dagitty

- DAGitty: www.dagitty.net

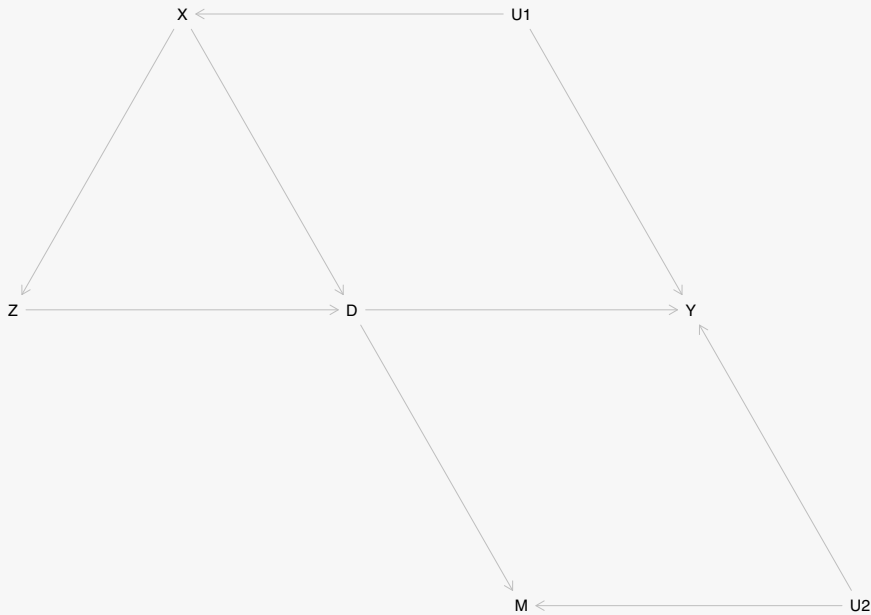
```
library(dagitty)
```

```
g <- dagitty('dag {  
  X [pos="1,-1.5"]  
  Y [pos="4,0"]  
  Z [pos="0,0"]  
  M [pos="3,1.5"]  
  D [pos="2,0"]  
  U1 [pos="3,-1.5"]  
  U2 [pos="5,1.5"]  
  
  X -> Z -> D -> Y  
  X -> D -> M  
  M <- U2 -> Y  
  X <- U1 -> Y  
}')
```

```
latents(g) <- c("U1", "U2")
```

R package: dagitty

`plot(g)`



R package: dagitty

```
parents(g, "D")
```

```
## [1] "X" "Z"
```

```
ancestors(g, "D")
```

```
## [1] "D" "Z" "X" "U1"
```

```
children(g, "D")
```

```
## [1] "M" "Y"
```

```
descendants(g, "D")
```

```
## [1] "D" "Y" "M"
```


R package: dagitty

```
paths(g, "D", "Y")$paths
```

```
## [1] "D -> M <- U2 -> Y"      "D -> Y"      "D <- X <- U1 -> Y"
```

```
## [4] "D <- Z <- X <- U1 -> Y"
```

```
paths(g, "D", "Y", directed = TRUE)$paths # only causal path(s)
```

```
## [1] "D -> Y"
```

R package: dagitty

```
dseparated(g, "Z", "D", c("X")) # because of Z -> D
```

```
## [1] FALSE
```

```
dseparated(g, "Z", "M", c("D"))
```

```
## [1] TRUE
```

```
impliedConditionalIndependencies(g)
```

$+ latent(g) \leftarrow c(V_1, V_2)$

```
## M _||_ X | D
```

~~V_1~~ , ~~V_2~~

```
## M _||_ Z | D
```

```
## Y _||_ Z | D, X
```

R package: dagitty

```
dseparated(g, "Z", "D", c("X")) # because of Z -> D
```

```
## [1] FALSE
```

```
dseparated(g, "Z", "M", c("D"))
```

```
## [1] TRUE
```

```
impliedConditionalIndependencies(g)
```

```
## M _||_ X | D
```

```
## M _||_ Z | D
```

```
## Y _||_ Z | D, X
```

R package: dagitty

```
adjustmentSets(g, "D", "Y", type="minimal")
```

```
## { X }
```

Caveat: adjustmentSets may include unobserved variables

which we cannot actually condition on.

```
S = adjustmentSets(g, "D", "Y", type="all")
```

```
S[!grepl("U1|U2", S)]
```

```
## { X } /
```

```
## { X, Z }
```

Note that this implements a slightly more general criterion

(sometimes it may contain descendants)

$$P(Y(d)) = \sum_m \int_t P(M_i=m | T_i=t) P(Y_i=t' | M_i=m) P(T_i=t)$$

* Frontdoor Criterion

