

Defining causal effects

STSCI / INFO / ILRST 3900: Causal Inference
Fall 2024

29 Aug 2024

Learning goals for today

By the end of class, you will be able to

- ▶ explain the fundamental problem of causal inference and the need for causal arguments
- ▶ define potential outcomes
- ▶ recall mathematical concepts from probability
 - ▶ random variables
 - ▶ expectation
 - ▶ conditional expectation

Causal claims hinge on arguments, not just data



Left photo: By Fernando Frazão/Agência Brasil - http://agenciabrasil.ebc.com.br/sites/_agenciabrasil2013/files/fotos/1035034-_mg_0802_04.08.16.jpg, CC BY 3.0, <https://commons.wikimedia.org/w/index.php?curid=50548410>
Right photo: By Agencia Brasil Fotografias - EUA levam ouro na ginástica artística feminina; Brasil fica em 8 lugar, CC BY 2.0, <https://commons.wikimedia.org/w/index.php?curid=50584648>

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- ▶ Swinging on the uneven bars causes a person to win a gold medal.

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What do we mean when we say “cause”?

Causal claims hinge on arguments, not just data

		Do you win gold if you:		Causal effect
		Swing	Do not swing	of swinging
Simone Biles		Yes (1)	?	?
	Sam	?	No (0)	?

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Sam	No (0)	No (0)		0



Fundamental problem of causal inference

Holland 1986

Descriptive evidence



Person 1	lifespan	
Person 2		lifespan
Person 3	lifespan	
Person 4		lifespan
Person 5	lifespan	
Person 6	lifespan	
Person 7		lifespan
Person 8	lifespan	

Outcome
under
Mediterranean
diet

Outcome
under
prudent
diet

Fundamental problem of causal inference

Holland 1986

Descriptive evidence



Causal claim



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Person 2		lifespan
Person 3	lifespan	
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lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan

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Fundamental problem of causal inference

Holland 1986

Descriptive evidence



Person 1	lifespan	missing
Person 2	missing	lifespan
Person 3	lifespan	missing
Person 4	missing	lifespan
Person 5	lifespan	missing
Person 6	lifespan	missing
Person 7	missing	lifespan
Person 8	lifespan	missing

Outcome
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Causal claim



lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
lifespan	lifespan
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Descriptive evidence



Causal claim



Causal inference is a **missing data** problem

Person 1	lifespan	missing
Person 2	missing	lifespan
Person 3	lifespan	missing
Person 4	missing	lifespan
Person 5	lifespan	missing
Person 6	lifespan	missing
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Person 8	lifespan	missing

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lifespan	lifespan
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Mathematical notation: Potential outcomes¹

¹Capital letters and lowercase letters mean different things!

Mathematical notation: Potential outcomes¹

Y_i Outcome

Whether person i survived

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Mathematical notation: Potential outcomes¹

Y_i	Outcome	Whether person i survived
A_i	Treatment	Whether person i ate a Mediterranean diet

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Y_i	Outcome	Whether person i survived
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Y_i^a	Potential Outcome	Outcome person i would realize if assigned to treatment value a

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Examples:

If assigned prudent diet	If assigned mediterranean diet
Died	Survived

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$Y_{\text{Sam}} = \text{survived}$ We observe that Sam survived

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$A_{\text{Sam}} = \text{MedDiet}$ We observed that Sam ate a Mediterranean diet

$Y_{\text{Sam}}^{\text{MedDiet}} = \text{survived}$ If Sam had been assigned a Mediterranean diet he would have survived

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$Y_{\text{Sam}}^{\text{MedDiet}} = \text{survived}$ If Sam had been assigned a Mediterranean diet
he would have survived

$Y_{\text{Sam}}^{\text{PruDiet}} = \text{died}$ If Sam had been assigned a prudent diet
he would have died

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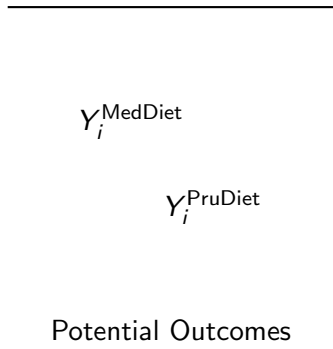
Practice

Using the slip of paper you received and the diet you follow, what is

- ▶ Y_i
- ▶ A_i
- ▶ Y_i^{MedDiet}
- ▶ Y_i^{PruDiet}

The consistency assumption

The consistency assumption



The consistency assumption



A square box with a horizontal top line and a vertical right line. Inside the box, the text Y_i^{MedDiet} is positioned higher and to the left, and Y_i^{PruDiet} is positioned lower and to the right.

Y_i^{MedDiet}

Y_i^{PruDiet}

Potential Outcomes



A square box with a horizontal top line and a vertical left line. Inside the box, the text Y_i is centered.

Y_i

Factual Outcomes

The consistency assumption

Consistency Assumption

$$Y_i^{A_i} = Y_i$$

Y_i^{MedDiet}

Y_i^{PruDiet}

Potential Outcomes

Y_i

Factual Outcomes

Notation: Expectation operator

A **random variable**² Y is a process which assigns outcomes to different cases by chance

²Dropping the sub-script means we are considering a random variable

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The **expectation operator** $E(\cdot)$ denotes the population mean

- ▶ The average if we had an infinite amount of data
- ▶ If the population of interest has n individuals

$$E(Y^a) = \frac{1}{n} \sum_{i=1}^n Y_i^a = (Y_1^a + Y_2^a + \dots Y_n^a)/n$$

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A **conditional expectation** is denoted with a vertical bar

$$E(Y \mid A = a) = \frac{1}{n_a} \sum_{i:A_i=a} Y_i$$

²Dropping the sub-script means we are considering a random variable

Practice: How would you say this in English?

We might wonder how a person's earnings relate to whether they hold a college degree

1. $E(\text{Earnings} \mid \text{Degree} = \text{TRUE}) > E(\text{Earnings} \mid \text{Degree} = \text{FALSE})$

2. $E(\text{Earnings}^{\text{Degree}=\text{TRUE}}) > E(\text{Earnings}^{\text{Degree}=\text{FALSE}})$

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► Average earnings are higher among those with college degrees

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► Average earnings are higher among those with college degrees

2. $E(\text{Earnings}^{\text{Degree}=\text{TRUE}}) > E(\text{Earnings}^{\text{Degree}=\text{FALSE}})$

► On average, a degree causes higher earnings

Practice:

1. On average, individuals who eat a Mediterranean diet survive more/less than those who eat a prudent diet:
2. On average, eating a Mediterranean diet causes people to survive more/less

Practice:

1. On average, students who do the homework learn more than those who don't
2. On average, doing the homework causes more learning

Practice:

1. On average, students who do the homework learn more than those who don't

$$E(\text{Learning} \mid \text{HW} = \text{TRUE}) > E(\text{Learning} \mid \text{HW} = \text{FALSE})$$

2. On average, doing the homework causes more learning

$$E(\text{Learning}^{\text{HW}=\text{TRUE}}) > E(\text{Learning}^{\text{HW}=\text{FALSE}})$$

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You can now

- ▶ Read Chapter 1 of [Hernán and Robins 2020](#)
- ▶ Complete survey on canvas