Data 102 Lecture 14:

Causal inference II

What is causality?

We mean different things when we say "cause"

The soccer ball moved because I kicked it.

My girlfriend broke up with me because I wasn't spending enough time with her.

I am who I am today because of my mum.

Gravity causes objects to fall to Earth.

Down Syndrome is caused by an extra copy of chromosome 21.

Humans are causing climate change.

Barbarian invasions caused the Roman Empire to fall.

All of form

A causes B

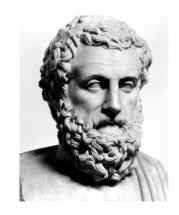
But meanings slightly different

Aristotle's definition

A cause is an answer to a "why" question

Four types of causes

- The **material** cause: "that out of which", e.g., the bronze of a statue.
- The **formal** cause: "the form", "the account of what-it-is-to-be", e.g., the shape of a statue.
- The **efficient** cause: "the primary source of the change or rest", e.g., the artisan, the art of bronze-casting the statue, the man who gives advice.
- The **final** cause: "the end, that for the sake of which a thing is done", e.g., health is the end of walking, losing weight, purging, drugs, and surgical tools.

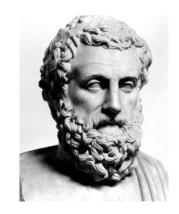


Modern scientific use focuses on **efficient** cause

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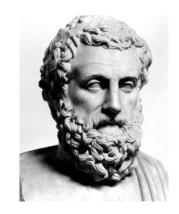


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Other differences in meaning...

Deterministic cause vs probabilistic cause

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Instance vs class

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Single cause vs multiple causes

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Simple action/event vs compound action/event

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Each domain of knowledge has their own

conventions / nuances

Death certificate

	f events-disea piratory arrest	SE OF DEATH (See instructions and ex- ses, injuries, or complicationsthat directly caused the death. DO NOT enter , or ventricular fibrillation without showing the etiology. DO NOT ABBREVIAT	terminal events such as	Approximate intervat: Onset to death
IMMEDIATE CAUSE (Final disease or condition	b. Pnet Due to (or c. COV	te respiratory distress syndrome or as a consequence of): umonia r as a consequence of): (ID-19 or as a consequence of):		2 days 10 days 10 days
PART II. Enter other significan	t conditions co	ntributing to death but not resulting in the underlying cause given in PART I.	33. WAS AN AUTOPSY PERFORMED? ☐ Yes ■ No 34. WERE AUTOPSY FINDINGS AVAILABLE THE CAUSE OF DEATH? ☐ Yes ☐ No	E TO COMPLETE
35. DID TOBACCO USE COLDEATH? Yes Probably No Unknown	NTRIBUTE TO	36. IF FEMALE: ■ Not pregnant within past year □ Pregnant at time of death □ Not pregnant, but pregnant within 42 days of death □ Not pregnant, but pregnant 43 days to 1 year before death □ Unknown if pregnant within the past year	37. MANNER OF DEATH Natural Homicide Accident Pending Invest Suicide Could not be d	

Takeaways

We have an intuitive sense of what causality means, but it's hard to make it precise.

In everyday language, our use of the word "cause" has different meanings

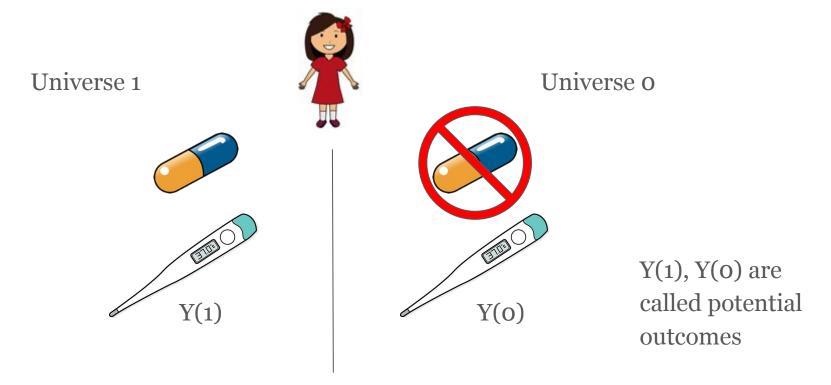
Different domains has its own nuances and conventions about how they define causes

A lot of ongoing philosophical debate

Lecture 14 overview

- Formalizing causality using counterfactuals
- Randomized experiments
 - Assumptions
 - Examples
 - What is the randomness
- Inference for randomized experiments
- Complications with randomized experiments

We define causality using counterfactuals



Individual treatment effect is Y(1) - Y(0)

Fundamental problem of causal inference:

We only see one of the two potential outcomes, so the individual treatment effect is unidentified

Ronald Fisher's insight

Do a randomized experiment!

- Consider multiple individuals (units) at the same time
- Use randomization as a basis for inference



The Science Table of a randomized experiment



i	$Y_i(1)$	$Y_i(0)$
1	$Y_1(1)$	$Y_1(0)$
2	$Y_2(1)$	$Y_2(0)$
:		:
n	$Y_n(1)$	$Y_n(0)$

The Science Table of a randomized experiment



i	$Y_i(1)$	$Y_i(0)$
1		$Y_1(0)$
2	$Y_2(1)$?
:		:
n	. .	$Y_n(0)$

Some formalism / definitions

 Z_i is the treatment assignment for unit i

 In the basic setting, treatment is binary, but can consider treatments with more levels

Y_i(1) and Y_i(0) are the potential outcomes of unit i

We observe $Y_{i,obs} = Y_i(Z_i)$

In order for Science Table to make sense...

Need the **Stable Unit Treatment Value Assumption** (SUTVA)

- 1. No interference: Potential outcomes for unit i do not depend on treatment assigned to other units
- 2. There are no other version of the treatment

Examples of experiments

- 1. We are interested in the effect of participating in a job training program or not on employment and wage. The intervention is participating in a job training program.
- 2. Gerber et al. (2008) were interested in the effect of different get-out-to-vote (GOTV) messages on the voting behavior. The intervention is different GOTV messages.
- 3. 2008 Oregon Health Insurance Experiment: Investigated the effect of health insurance on health and economic outcomes. The intervention was lottery Medicaid enrollment via lottery.
- 4. Technology companies are typically interested in how different versions of an app affects user engagement.

SUTVA not always reasonable...

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- 2. Gerber et al. (2008) were interested in the effect of different get-out-to-vote (GOTV) messages on the voting behavior. The intervention is different GOTV messages. People in same household will receive the same message.
- 3. 2008 Oregon Health Insurance Experiment: Investigated the effect of health insurance on health and economic outcomes. The intervention was lottery Medicaid enrollment via lottery.
- 4. Technology companies are typically interested in how different versions of an app affects user engagement. If User A gets affected by treatment, then he/she will interact more/less with their friends.

The randomization scheme

Determined by the people running the experiment.

Typically, do complete randomization.

$$\mathbb{P}(Z=z) = \frac{1}{\binom{n}{n_1}}$$

Where $z = (z_1, ..., z_n)$ satisfies $\sum_{i=1}^n z_i = n_1$ and $\sum_{i=1}^n (1-z_i) = n_0$

I.e. we choose $\mathbf{n}_{_1}$ units uniformly without replacement to form our treatment group

Inference for randomized experiments

The Average Treatment Effect (ATE)

$$\tau = \frac{1}{n} \sum_{i=1}^{n} (Y_i(1) - Y_i(0))$$

This is the average of individual treatment effects

This is a fixed quantity. The potential outcomes are assumed to be fixed.

It is unidentifiable, can only estimate it.

The Neyman estimator (difference-in-means)

$$\hat{\tau} = \frac{1}{n_1} \sum_{Z_i = 1} Y_{i,obs} - \frac{1}{n_0} \sum_{Z_i = 0} Y_{i,obs}$$

Mean and variance for Neyman estimator

$$\mathbb{E}[\hat{\tau}] = \tau$$

$$\operatorname{Var}(\hat{\tau}) \le \frac{S_1^2}{n_1} + \frac{S_0^2}{n_0}$$

Where for k = 0, 1:
$$S_k^2 = \frac{1}{n_k - 1} \sum_{i=1}^{n_k} (Y_i(k) - \bar{Y}(k))^2$$

Confidence intervals for ATE

See whiteboard notes

Hypothesis testing for randomized experiments

Neyman's weak null H_{oN} : ATE = 0

Fisher's strong null H_{oF}: ITE $Y_i(1) - Y_i(0) = 0$ for each unit i

Much debate about which null hypothesis is more meaningful.

See whiteboard notes for more details

Fisher's exact test

See whiteboard notes

Data example: Racial discrimination in the labor market

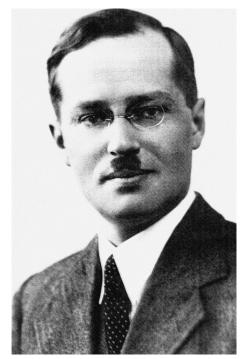
Bertrand, M. and Mullainathan, S. (2004) wanted to study whether there was racial discrimination in the labor market

Experiment: Create 4870 fake resumes, half of them with stereotypical white names, and half of them with stereotypical black names, sent them to potential employers

Treatment: Race

Outcome: Whether or not the resume received a call back.

Background: Neyman-Rubin framework



Jerzy Neyman (1894-1981)



Donald Rubin (1943-)

Summary about inference under Neyman-Rubin framework

Does not require modeling assumptions on the potential outcomes

Not completely assumption-free: SUTVA

This is a frequentist framework

Randomness comes from the experiment design, not from sampling

Causality is formulated in terms of instances (what happens to me), but is estimated in terms of a class (what happens on average to everyone)

Heterogeneity is unaccounted for (for now...)

Complications with randomized experiments

Compliance

Units may not comply with their assigned treatment

E.g. 1: In medical trials, patients sometimes decline their assignment treatment (stop taking their medication, or decline surgical procedure.)

E.g. 2. In the Oregon Health Insurance experiment, only 45% of lottery winners enrolled in Medicaid. In addition, 18% of lottery losers enrolled.

External validity

The ATE is defined with respect to the experiment population. This may not be representative of the larger population of interest.

E.g. Most polls in the 2016 US presidential election overestimated the support for Hillary Clinton. This polling error was due to the polled population being different from the voting population (for instance, Trump voters were less likely to answer the phone)

• This example is not exactly about causal inference, but this is the crux of the external validity problem

