

# Causality

POST 8000 – Foundations of Social Science Research for Public Policy

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## Goal for Today

*Introduce students to causality, and distinguishing causality from association.*

## The Problem, in Quotes

- “That correlation is not causation is perhaps the first thing that must be said.” - Barnard, 1982 (p. 387)
- “If statistics cannot relate cause effect, they add to the rhetoric.” - Smith, 1980 (p. 1000 [stylized by me])

# Associational Inference

A set of tools to understand how a response variable corresponds with some attribute.

Tools include:

- Probability distributions (conditional, joint)
- Correlation
- Regression(?)

“Associational inference consists of [estimates, tests, posterior distributions, etc.] about the associational parameters relating  $Y$  and  $A$  [from units in  $U$ ]. In this sense, associational inference is simply descriptive statistics.” - Holland, 1986 (p. 946)

# Probability Distributions

**Joint probability**, in the event  $A$  and  $B$  are independent from each other:

$$p(A, B) = p(A) * p(B)$$

**Conditional probability**, in the event that  $A$  depends on  $B$  having already occurred:

$$p(A | B) = \frac{p(A, B)}{p(B)}$$

## Correlation (via Pearson's $r$ )

$$\sum \frac{\left(\frac{x_i - \bar{x}}{s_x}\right)\left(\frac{y_i - \bar{y}}{s_y}\right)}{n - 1}$$

...where:

- $x_i, y_i$  = individual observations of  $x$  or  $y$ , respectively.
- $\bar{x}, \bar{y}$  = sample means of  $x$  and  $y$ , respectively.
- $s_x, s_y$  = sample standard deviations of  $x$  and  $y$ , respectively.
- $n$  = number of observations in the sample.

## Properties of Pearson's $r$

1. Pearson's  $r$  is symmetrical.
2. Pearson's  $r$  is bound between -1 and 1.
3. Pearson's  $r$  is standardized.

# Standardization

$$z = \frac{\text{Deviation from the mean}}{\text{Standard unit}}$$

The standard unit will vary, contingent on what you want.

- If you're working with just one random sample, it's the standard deviation.
- If you're comparing sample means across multiple random samples, it's the standard error.



# Standardization

Larger  $z$  values indicate greater difference from the mean.

- When  $z = 0$ , there is no deviation from the mean (obviously).

Standardization has a lot of cool properties you'll see through the semester.

- For now: it's a way to express a variable's scale.

# Causal Inference

Causal inference owes much to Rubin's "potential outcomes framework."



# The Problem in a Nutshell

An individual ( $i$ ) who is offered a treatment ( $Z_i = 1$ ) has two potential outcomes:

- An outcome to be revealed if treated ( $T_i = 1$ ):  $Y_i(T_i = 1|Z_i = 1)$
- An outcome to be revealed if *untreated* ( $T_i = 0$ ):  $Y_i(T_i = 0|Z_i = 1)$

This is a missing data problem of a kind.

- We can only observe one.
- No perfect counterfactuals.
- Unicorns don't exist.

## The Solution

For  $T_i = 0$  and  $T_i = 1$ , given both offered treatment ( $Z_i = 1$ ):

$$\text{Individual Treatment Effect for } i = Y_i(T_i = 1|Z_i = 1) - Y_i(T_i = 0|Z_i = 1)$$

Think in terms of population averages.

- Per Rubin, there is an important population parameter to estimate.
- Hence why we (and he) referred to it as “effect of the treatment on the treated.” (i.e. TOT)
- Also: the “average treatment effect” (i.e. ATE)

# The Importance of Random Assignment

Random assignment (to treatment/control) helps us with ATE because it's tough to imagine cases where ( $Z_i = 1$  and  $T_i = 0$ ).

- Per random assignment: participants assigned to treatment/control must be same on average in the population ("equal in expectation").
- i.e.  $E[Y_i(T_i = 0|Z_i = 1)]$  must be equal to  $E[Y_i(T_i = 0|Z_i = 0)]$

By substitution:

$$TOT = E[Y_i(T_i = 1|Z_i = 1)] - E[Y_i(T_i = 0|Z_i = 0)]$$

When unbiased, a difference in sample means is sufficient:

$$T\hat{O}T = \frac{\sum_{i=1}^{n_1} Y_i}{n_1} - \frac{\sum_{i=1}^{n_0} Y_i}{n_0}$$

## Some Other Important Assumptions

- Exogeneity (worth reiterating)
- Unit homogeneity
- Conditional independence
- SUTVA

# Criteria for Evaluating Causal Arguments

- Falsifiability
- Internal consistency
- Careful selection of DV
- Concreteness
- “Encompassibility” (sic)

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