

# Social Science Inquiry II

## Week 2: Course introduction, part II

Molly Offer-Westort

Department of Political Science,  
University of Chicago

Winter 2024

# Is social science *science*?

- ▶ In discussing methods for social science research, we should think about what social science is, and what its goals are.
- ▶ What is science?
- ▶ What is social science?
- ▶ Is social science *science*? Why?

# Definition of scientific research [KKV]

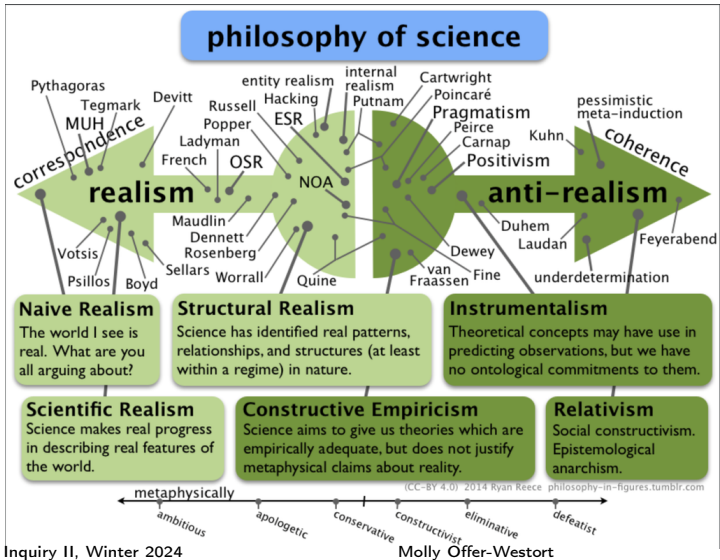
1. The goal is inference.
2. The procedures are public.
3. The conclusions are uncertain.
4. **The content is the method.**

# Definition of social science [KKV]

“Social science constitutes an attempt to make sense of social situations that we perceive as more or less complex.”

# Philosophy of science

- ▶ These terms do not have definitions that are universally agreed on.
- ▶ ...and there is a whole field on the philosophy of science.



# Approach

- ▶ We will focus on study that is empirical, where scientific knowledge is built based on observed, measured phenomena.
- ▶ There are other approaches.

# The content is the method.

- ▶ Claim: Research is science if it uses a scientific method.

# Path of research development

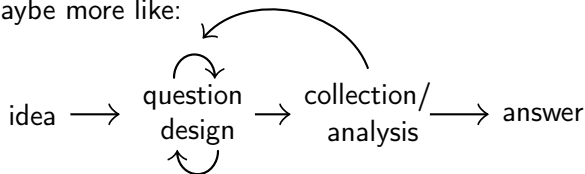
A common linear paradigm for a research project:

question → design → collection → analysis → answer

Tukey (1980) framing:



Maybe more like:





# Proposing theories [KKV]

- ▶ Propose theories that are *falsifiable*.
- ▶ If you want to be able to test this theory, it should have *observable implications*. (The more the better.)
- ▶ Clarity and precision in your theory is helpful.

# How does this inform how we think about data?

- ▶ Data is not something that is handed down to us in its complete and final form—we select and shape our data.

# Principles of data collection [KKV]

1. Record and report data generation process.
2. Collect data on as many observable implications as possible.
3. Maximize the validity of measurements.
4. Ensure that data-collection methods are reliable.
5. Data and analyses should be as replicable as possible.

# Observable implications

- ▶ What are these?
- ▶ How do observable implications relate to causal inference?

# Inference

- ▶ What is descriptive inference?
- ▶ What is causal inference?

# Elements of (descriptive/causal) inference [Holland]

- ▶ Define a population of interest.  $U$
- ▶ Determine a variable that is defined over this population, for which there is variation.  $A$
- ▶ Define response variable “of interest.”  $Y$

We can consider how  $A$  and  $Y$  vary together. Without further assumptions, this is just correlation.

What does it mean for something to cause something else?

# What does it mean for something to cause something else? [Rubin]

- ▶ Possibility for multiple values of treatment—for *each individual*. Treatment *could have been different*.  $S(u) = t$  OR  $S(u) = c$
- ▶ Some temporal element: outcome variable is realized after “exposure” to treatment variable.  $S(u) \rightarrow Y(u)$
- ▶ Conceptual possibility for multiple different versions of response variable—for *each individual*.  $Y_t(u)$  may not equal  $Y_c(u)$

Consider the individual *causal effect*

$$Y_t(u) - Y_c(u).$$



# No causation without manipulation [Holland]

- ▶ Causes are only things that could (hypothetically) be treatments in experiments.
- ▶ NOT attributes—e.g., race and biological sex can't plausibly be manipulated, because there is no meaningful *counterfactual*.
  - ▶ What we can manipulate: others' *perception* of race/sex (Bertrand and Mullainathan, 2004)
  - ▶ To the extent that sex/gender are social constructs, can we consider counterfactual socialization?

# Fundamental problem of causal inference

- ▶ We only see the response variable under one version of treatment.
- ▶ Why does this matter?

# Resolution (?) of the fundamental problem of causal inference

- ▶ Causal inference is not impossible.
- ▶ But making causal inferences without making assumptions IS impossible.
- ▶ So we need to depend on some assumptions.
- ▶ The crux: how plausible are these assumptions?

# Some special cases of causal inference

- ▶ Temporal stability
- ▶ Causal transience
- ▶ Unit homogeneity
- ▶ Constant causal effect
- ▶ Independence

# Temporal stability and causal transience

- ▶ The value of response will not change based on *when* you apply treatment to an observation.
- ▶ The value of response will not change if you had at some point previously applied a different treatment to an observation.

# Unit homogeneity

- ▶ The value of response under a given treatment is the same for two observations; and the value of response for a different treatment is also the same.
- ▶ You can observe response on the comparable units under one of each version of treatment.

# Constant causal effect

- ▶ The difference between the value of the response variable under one version of treatment as compared to another version of treatment is the same for *every* observation.
- ▶ How does this relate to the unit homogeneity assumption?
- ▶ Does this allow us to back out the value of the causal effect, without other assumptions?

# Independence

- ▶ There are a “large” number of observations.
- ▶ On average, the units that are assigned one version of treatment look like the units that receive another version of treatment.
- ▶ (What does  $E[Y_s|S = t]$  mean?)



# The special role of experiments

- ▶ In experiments, treatment is randomly assigned by the researcher.
- ▶ So we know independence holds by design.
- ▶ We can compare groups and get the *average* causal effect.
- ▶ (When can we also get the *individual* causal effect?)

# References I

- Bertrand, M. and Mullainathan, S. (2004). Are emily and greg more employable than lakisha and jamal? a field experiment on labor market discrimination. American economic review, 94(4):991–1013.
- Tukey, J. W. (1980). We need both exploratory and confirmatory. The American Statistician, 34(1):23–25.