

# Social Science Inquiry II

## Week 1: Course introduction, part II

Molly Offer-Westort

Department of Political Science,  
University of Chicago

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# 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]

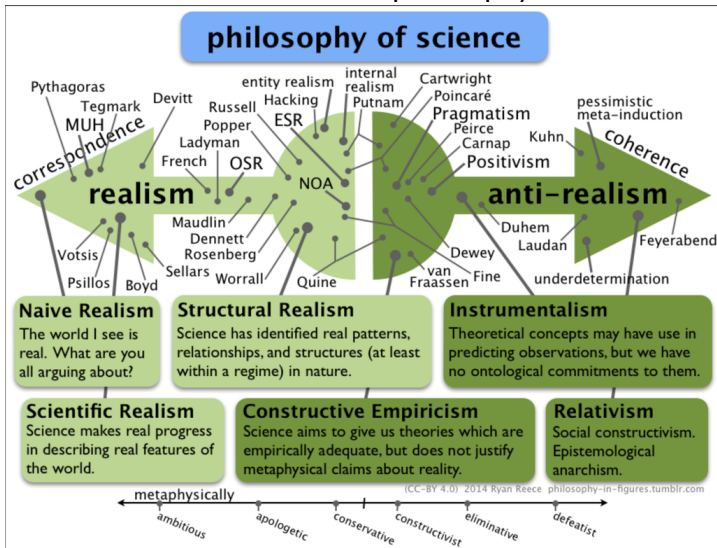
1. The goal is inference.
2. The procedures are public.
3. The conclusions are uncertain.
4. **The content is the method.**

# 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

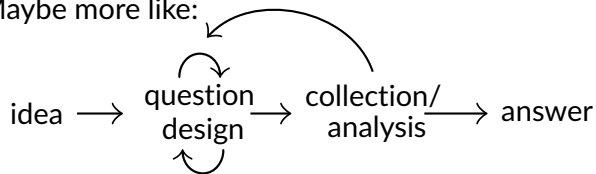
A common linear paradigm for a research project:

question  $\rightarrow$  design  $\rightarrow$  collection  $\rightarrow$  analysis  $\rightarrow$  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.