

# The Philosophy of Causal Inference

## Cause versus causal effect.

- We need a “treatment.”
- We need a comparison.

## Potential outcomes framework.

- $Y_i(0)$  and  $Y_i(1)$ .

## Causal inference is a missing data problem. (The “Fundamental Problem of Causal Inference.”)

- We never observe *both* potential outcomes.
- We observe  $Y_i(0|T_i = 0)$ ,  $Y_i(1|T_i = 1)$ ,  
but not  $Y_i(1|T_i = 0)$ ,  $Y_i(0|T_i = 1)$ .

# The Practice of Causal Inference

**Two assumptions.**

- 1 SUTVA.
- 2 Ignorability.

**Randomization as the basis for valid inference (Fisher).**

**Without randomization we need conditional independence.**

- $Y_i(0), Y_i(1) \perp T_i \mid X_i$

**How important is the mode of inference?**

# Lessons and Issues for Social Science

**This applies to qualitative and quantitative research.**

**Treatment assignment.**

- “Design-based” research.

**Assumptions: testable and untestable.**

**“Regression is evil” (and other stuff is better).**

**Can we have principled methods?**

**Should we abandon “fundamentally unanswerable questions”?**

# Going Beyond Causal Inference

Causal inference is a powerful and compelling framework for thinking about this stuff.

**What is the role for descriptive work?**

**What is the role for data mining?**

**What is the role for theory?**

**Are causal effects always interesting??**