PMAP 8131 Applied Research Methods II Causality

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Outline

- Epistemology
- Causation
- Validity
 - Threats to validity

Table of Contents

- Epistemology
- 2 Causation
- Validity
 - Threats to validity

- Popper's falsification principle
- Kuhn's paradigm shift
- Lakatos research paradigm

Popper's falsification principle

• Logical asymmetry of verification and falsification

Modus ponens

• Premise I: All dogs bark.

• Premise II: Tyson is a dog.

Conclusion: Tyson barks.

Modus tollens

• Premise I: All dogs bark.

• Premise II: Tyson doesn't bark.

• Conclusion: Tyson is not a dog.

Modus ponens

- 1. $D \rightarrow B$
- 2. D
- 3. B = Implication: 1,2

Modus tollens

- 1. $D \rightarrow B$
- 2. ¬*B*
- 3. $\neg D$ = Implication: 1,2

- Logical asymmetry of verification and falsification
 - One negative obs is enough to invalidate a theory
 - No number of positive obs would ever validate a theory
- Demarcationist stance
 - Falsification demarcates science from pseudo-science
 - Difference between statements is logical, not empirical

Duhem-Quine Thesis

1.
$$(T1 \wedge T2 \wedge T3) \rightarrow O$$

$$2. \neg O$$

3.
$$\neg (T1 \land T2 \land T3)$$

Kuhn's scientific revolutions: Popper + Duhem-Quine

- Local vs global confutations
 - Theories are bundles of propositions $T = \{T_1, \dots, T_n\}$
 - Local confutations do not prove a theory false
- Normal science vs revolutionary science
 - Normal science is puzzle-solving of local confutations and never discovers anything new

Lakatos' research paradigms: Marx + Hegel

- Science as "struggle of theories"
 - Theories: $T^1 = \{T_1^1, \dots, T_n^1\}, \dots, T^n = \{T_1^n, \dots, T_n^n\}$
 - A progressive/regressive theory generates more/less successes, on average, than competitor theories
- Theories as "paradigms"
 - Sub-theories layer up onto theories, only upper layer theories brought to the fore of evidence

Falsification: Regression example

- **Hypothesis**. Male-female math achievement gap due to institutional features (X_i) , does not reflect genetics
- Experimentum crucis. Sequential regression models:

$$MATH_i = \beta_0 + \beta_1 Female_i + \epsilon_1$$
 (1)

$$MATH_i = \beta_0 + \beta_1 Female_i + \beta_2 X_i + \epsilon_1$$
 (2)

Falsification: Regression example

- **Hypothesis**. Male-female math achievement gap due to institutional features (X_i) , does not reflect genetics
- Experimentum crucis. Sequential regression models:

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 (1)

$$MATH_i = \beta_0 + \beta_1 Female_i + \beta_2 X_i + \epsilon_1$$
 (2)

• **Decision rule**: Reject hypothesis if β_1 significant in model 1), but insignificant in model 2)



Falsification: Regression example (Guiso et al., 2008)

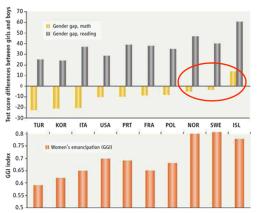


Figure: Female-male gap reverts in countries with greater gender parity

Human subjects

 Act in such a way that you treat humanity, whether in your own person or in the person of another, always at the same time as an end and never simply as a means (Immanuel Kant)

Human subjects: Violations

- Violence on experimentees
- Framing of questions
- Omission of information
- Interaction with experimentees



Data management

 There are three kinds of lies: Lies, Damned Lies, and Statistics (Mark Twain)

Data management: Violations

- Addition or removal of observations
- "P-hacking"
- Framing of findings
- Uncommented code!!!

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Causation

- Correlation
- 2 Time-order
- Non-spuriousness

Correlation

$$\rho(X,Y) = \frac{\sigma_{XY}}{\sigma_X \sigma_Y} \approx \frac{COV_{XY}}{SD_X SD_Y} \approx \frac{\sum (X - \overline{X})(Y - \overline{Y})}{SD_X SD_Y}$$

- Correlation: Examples
 - Size of left and right foot
 - Watching football and earnings
 - Leap year and Olympic Games

- Non-spuriousness
 - Relationship between X and Y spurious to Z
 - Non-spuriousness: Examples
 - Air conditioning and IQ test score (Z: Temperature)
 - Suicide and Reddit engagement (Z: Depression)
 - Car usage and individual longevity (Z: Income)

- Time-order
 - X comes before Y (Beware: reverse causation!)

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Validity

Experiment

- Subject features
- Experimental conditions
- Experimental stimulus

Validity

- Internal validity
 - Equality of treated and controls

- External validity
 - Equality of experimentees and outer population

Validity

- Internal validity
 - Equality of treated and controls
 - Observables and unobservables features of treatments that affect the outcome match features of controls
- External validity
 - Equality of experimentees and outer population
 - Experimental conditions and stimulus match inputs in the real world

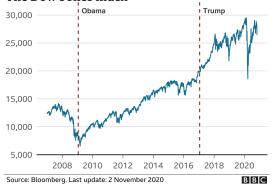
- Threats to internal validity
 - History
 - Maturation
 - Instrumentation
 - Regression to the mean
 - Self-selection
 - Differential mortality

- Threats to external validity
 - Measuring
 - Testing
 - General equilibrium effects
 - Multiple treatments
 - Self-selection

History. Treatment effect are attributable to the passing of time.

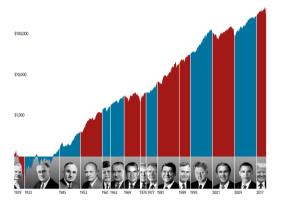
- Example: Federal policies and economic growth rates
 - Economy follows own short- and long-run trends
 - Political cycle overlaps with economy, does not cause it

Stock market growth and jitters: The Dow Jones index



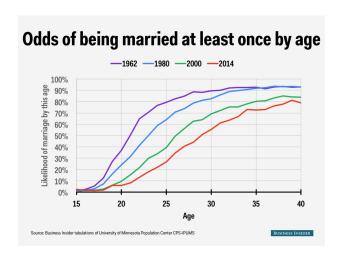
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GROWTH OF \$100
FAMA/FRENCH TOTAL US MARKET RESEARCH INDEX
March 4, 1929–June 30, 2020



Maturation. Treatment effect are attributable to the aging of subjects.

- Example: Welfare programs and marital outcomes
 - People more likely to marry and have kids as they age



Instrumentation. The effect is measured with instrumental and nonrandom error.

- Example: Domain knowledge tests
 - Test measures test-taking ability, not domain knowledge

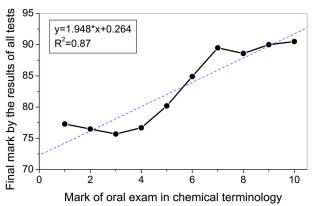


Figure: Correlation of chemistry test scores and other tests (Gryshchenko et al., 2021)



Regression to the mean. Treatment is observed at a point below or above their average value.

- Example: Job training programs
 - Trainees likely to get back to workforce anyways

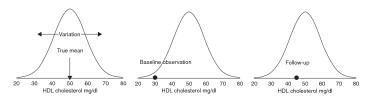


Figure: Regression to baseline cholesterol (Barnett et al., 2005)

Self-selection. Features of experiment drive selection into treatment group.

- Example: Immigrant earnings
 - Migrants systematically different from those who stay

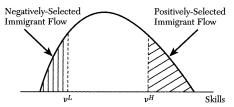


Figure: Immigrant self-selection (Borjas, 1994)

Differential mortality. Treatment and control exit experiment at different and nonrandom rates.

- Example: Welfare programs
 - Burdensome paperwork forfeits unmotivated treatments

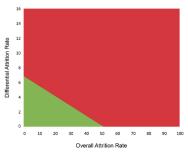


Figure: Maximum differential attrition (Mathematica, 2015)



Measuring. Reaction to measuring affects the outcome.

- Example: Observing experimentees
 - Behaviors change when we know we are being measured

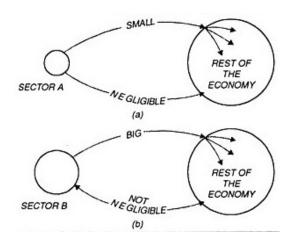
Testing. Reaction to testing affects the outcome.

- Example: Pre-testing
 - Pre-testing increases test-taking skills

Group	Pretest	Treatment	Posttest
1	O_1	X	O_2
2	O_1	_	O_2
Time ▶			

General equilibrium effects. Generalization of treatment to population forfeits local effects.

- Example I: The "paradox of schooling"
 - If I get an A, I stand out. If everyone gets an A, no one stands out
- Example II: Government spending
 - If I get a subsidy, I am better off. If everyone gets a subsidy, inflation makes us all worse off



Multiple treatments. Treatment effect does not generalize if treatment is administered jointly with a second treatment.

- Example: Psychological counseling
 - If counseling involves commutes, treatment effect is the joint effect of counseling and taking time off work

Self-selection. Features of experiment drive selection into experimental group.

- Example: Student evaluations of professors
 - Students leave evaluations when very happy or unhappy