## Causality and Validity EDP 612 Week 7

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Introduction to Research

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#### **Experiments and Causation**

#### Cause

- Variable that produces an effect or result
- Most causes are inus -

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A cause is an insufficient (i)
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but non-redundant (**n**)

```
part of an unnecessary (u) but
```

```
sufficient condition (s)
```

- A given event may have many different causes
- Many factors are required for an effect to occur, but they can rarely be fully known and how they relate to one another

#### Effect

- Difference between what did happen and what would have happened
- This reasoning generally requires a counterfactual

#### Counterfactual

- Knowledge of what would have happened in the absence of a suspected causal agent
  - Physically impossible
  - Impossible to simultaneously receive and not receive a treatment
  - Therefore, the central task of all cause-probing research is to approximate the physically impossible counterfactual

#### **Causal Relationships**

A causal relationship requires three conditions

- 1. Cause preceded effect (temporal precedence)
- 2. Cause and effect covary
- 3. No other plausible alternative explanations can account for a causal relationship

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#### Cause, Effect, and Causal Relationships

- In experiments
  - Presumed causes are manipulated to observe their effect
  - Variability in cause related to variation in an effect
  - Elements of design and extra-study knowledge are used to account for and reduce the plausibility of alternative explanations

#### Causation, Correlation, and Confounds

- Correlation does not prove causation
- Correlations do not meet the first premise of causal logic (temporal precedence)
- Such relationships are often due to a third variable (i.e., a confound)

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#### **Manipulable and Nonmanipulable Causes**

- Experiments involve causal agents that can be manipulated
- Nonmanipulable causes (e.g., ethnicity, gender) cannot be causes in experiments because they cannot be deliberately varied

#### **Causal Description and Causal Explanation**

- Causal description. identifying that a causal relationship exists between A and B
- Molar causation. the overall relationship between a treatment package and its effects
- Causal explanation. explaining how A causes B
- Molecular causation. knowing which parts of a treatment are responsible for which parts of an effect

#### **Causal Models**

A. Causal description (direct)



B. Causal explanation (indirect)



C. Causal explanation (direct and indirect)



#### **Causal Models**

A. Moderator model



B. Mediator model



#### **Modern Descriptions of Experiments**

#### Randomized Experiment

- Units are assigned to conditions randomly
- Randomly assigned units are probabilistically equivalent based on expectancy (if certain conditions are met)
- Under the appropriate conditions, randomized experiments provide unbiased estimates of an effect

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#### **Quasi-Experiment**

- Shares all features of randomized experiments except assignment
- Assignment to conditions occurs by self-selection
- Greater emphasis on enumerating and ruling out alternative explanations
  - $\circ \ \ldots$  through logic and reasoning, design, and measurement

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#### **Natural Experiment**

- Naturally-occurring contrast between a treatment and comparison condition
- Typically concern nonmanipulable causes
- Requires constructing a counterfactual rather than manipulating one

#### **Nonexperimental Designs**

- Often called correlational or passive designs (i.e., cross-sectional)
- Statistical controls often used in place of structural design elements
- Generally do not support strong causal inferences

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#### **Experiments and the Generalization of Causal Connections**

# Most Experiments are Local but have General Aspirations

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- Most experiments are localized
- Limited samples of utos

units (**u**)

treatments (t)

observations (o)

settings (s)

• What Campbell labeled local molar causal validity

#### **Construct Validity: Causal Generalization as Representation**

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• Premised on generalizing from particular sampled instances of units, treatments, observations, and settings to the abstract, higher order constructs that sampled instances represent

## **External Validity: Causal Generalization as Extrapolation**

- Inferring a causal relationship to unsampled units, treatments, observations, and settings from sampled instances
- Enhanced when probability sampling methods are used
  - Broad to narrow
  - Narrow to broad

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#### Approaches to Making Causal Generalizations

- Sampling
- Probabilistic
- Heterogeneous instances
- Purposive
- Grounded theory
- Surface similarity
- Ruling out irrelevancies
- Making discrimination
- Interpolation and extrapolation
- Casual explanation

## Statistical Conclusion Validity and Internal Validity

## Validity

- Approximate truthfulness of correctness of an inference
- Not an all or none, either or, condition, rather a matter of degree
- Efforts to increase one type of validity often reduce others

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#### **Statistical Conclusion Validity**

Validity of inferences about the covariation between treatment (cause) and outcome (effect)

## Internal Validity

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Validity of inferences about whether observed covariation between A (treatment/cause) and B (outcome/effect) reflects a causal relationship from A to B as those variables were manipulated or measured

#### **Construct Validity**

Validity of inferences about the higher order constructs that represent sampling particulars



## **External Validity**

Validity of inferences about whether a cause-effect relationship holds over variations in units, treatments, observations, and settings

#### **Threats to Validity**

- Reasons why an inference may be partly or wholly incorrect
- Design controls can be used to reduce many validity threats, but not in all instances
- Threats to validity are generally context-dependent

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#### **Internal Validity**

- Inferences about whether the observed covariation between A and B reflects a causal relationship from A to B in the form in which the variables were manipulated or measured
- In most cause-probing studies, internal validity is the primary focus

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## Threats to Internal Validity (1/2)

- Ambiguous temporal precedence. Lack of clarity about which variable occurred first may yield confusion about which variable is the cause and which is the effect
- Selection. Systematic differences over conditions in respondent characteristics that could also cause the observed effect
- History. Events occurring concurrently with treatment that could cause the observed effect
- Maturation. Naturally occurring changes over time that could be confused with a treatment effect
- **Regression**. When units are selected for their extreme scores, they will often have less extreme scores on other variables, an occurrence that can be confused with a treatment effect

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## Threats to Internal Validity (2/2)

- Attrition. Loss of respondents to treatment or measurement can produce counterfactual effects if that loss is systematically correlated with conditions
- **Testing**. Exposure to a test can affect test scores on subsequent exposures to that test, an occurrence that can be confused with a treatment effect
- Instrumentation. The nature of a measure may change over time or conditions in a way that could be confused with a treatment effect
- Additive and interactive threats. The impact of a threat can be added to that of another threat or may depend on the level of another threat

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### **Estimating Internal Validity in Experiments**

- By definition randomized experiments eliminate selection through random assignment to conditions
- Most other threats are (should be) probabilistically distributed as well

### **Estimating Internal Validity in Experiments**

- Only two likely validity threats (typically) arise from experiments
  - Attrition
  - Testing

#### **Estimating Internal Validity in Quasi-Experime**

- Differences between groups tend to be more systematic than random
- All threats should be made explicit and then ruled out one by one
- Once identified, threats can be systematically examined

#### That's it!

Any questions?