

# Week 11 Lecture - Experimental Confounds

Undergraduate Research Methods in Psychology

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## **1** Learning Objectives

#### 1.1 Textbook Objectives

- Interrogate a study and decide whether it rules out twelve potential threats to internal validity.
- Describe how researchers can design studies to prevent internal validity threats.
- Interrogate an experiment with a null result to decide whether the study design obscured an effect or whether there is truly no effect to find.
- Describe how researchers can design studies to minimize possible obscuring factors.

#### 1.2 Professor's Objectives

- Describe the importance of comparison groups in ruling out threats to validity
- Explain an example for each type of validity threat

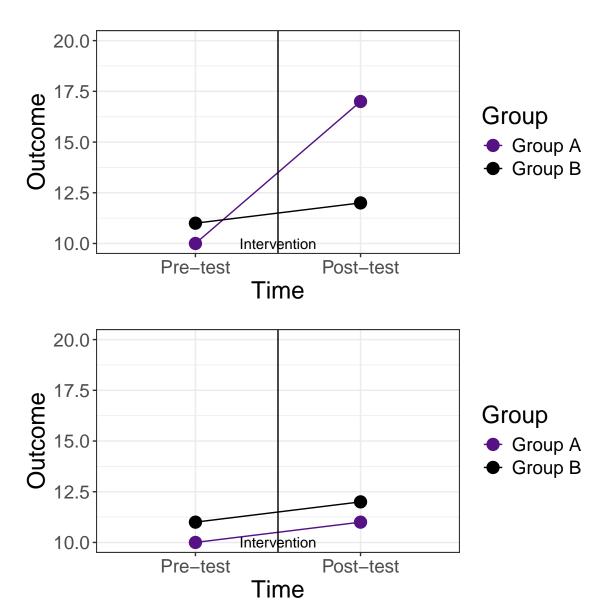
# **2** Chapter Overview

#### 2.1 Chapter Overview

•	Because of the	required to investigate cau	sal claims, we must	
	be cautious of a number of internal	validity threats		
•	We may have to be especially mind our finding are "real". Design flaws		in order to ensure les.	
•	Failure to design around these threat lidity, and therefore, may hurt our cl		r	_va-

# **3** More Internal Validity Threats

#### **3.1** Working Example



#### 3.2 Overview

- We have already discussed \_\_\_\_\_\_ effects, design confounds, and order effects as well as how they impact the \_\_\_\_\_ -subjects and within-subjects designs differently
- These all result from a lack of random
   not being implemented

• However, there are many other threats to also be concerned with...

## 3.3 For One-group & Pretest/Posttest Designs

• Some threats are a particular concern in our between-subjects designs:

3.3.1	Maturation	
•	Certain behaviors may simply this change as being "spontaneo	by themselves - we may describe ous" and unexpected
•	This is change that isthereof	explainable by our intervention or lack-
•	cause we have a comparison the maturation.	roup! Perhaps there is a maturation effect, but be- on group, we can see if there is a difference despite t accurately demonstrated with a pretest/posttest of change over
3.3.2	History	
•	An effect may occur due to som change in our groups.	ethat creates an unexpected
•	However, to be a true history threa in a outsized manner on only on	
•	same event, then the effect for the differences between - However, if the event some	
3.3.3	Regression	
•	This threat revolves around regreextreme scores tend to naturally	ssion to the, where, over time, converge towards the central tendency of the data

Thus, extremely high scores naturally tre- low scores naturally trend	end downward as time passes, and extremely
<del>-</del>	groups! (Notice a trend?) start, and a difference in trend, we know an effect above and beyond the effects of
3.3.4 Differences Between Maturation, H	istory, & Regression
• Maturation threats deal with spontaneo	us and unexplainable change in behavior
History threats are due to a tal influence	event or know outside environmen-
<ul> <li>Regression threats come form naturally the mean or center of a scale.</li> </ul>	scores converging on
3.3.5 Attrition	
<ul> <li>Attrition occurs whenever we have son sample (may also be known as "mortal</li> </ul>	
<ul> <li>If attrition happens among all groups ar cern is low.</li> </ul>	of people, then con-
<ul> <li>But, If there is high attrition among one that could confound</li> </ul>	e of the groups, we have a differential effect
<ul> <li>Prevention:         <ul> <li>It is often impossible to fully predefeating</li> <li>(because research is</li> </ul> </li> </ul>	vent people from not completing the study !)
<ul> <li>Instead, we may opt to data corresponding to those who of</li> </ul>	delete, or completely remove the dropped out.
3.3.6 Testing	
<ul> <li>As we learned with if they take an assessment more than a natural growth.</li> <li>Prevention:</li> </ul>	_effects, participants may grow more skilled once, not due to intervention, but just due to
i Tovontion.	

<ul> <li>Don't use a threats more dif</li> </ul>		lly, this makes inve	stigating the other
<ul> <li>Use two differer</li> </ul>	nt, but equivalent forms of ne same thing? $ ightarrow$	the same test! But	t how do we know
3.3.7 Instrumentation			
<ul> <li>This occurs when so over time.</li> </ul>	omething about the	ins	strument changes
• Put another way, our	measurement is not beha	ving	<u> </u> !
over time if it is - We may use cor	nly a post-test design. The only used once! nstruct validity statistics (e.gability of the measures at b		can't change $lpha$ ) to assess
<ul> <li>We may counted</li> </ul>	rbalance order of two	fc	orms
3.3.8 Combined Threats	5		
<ul> <li>Our selection effects here, or even multiple</li> </ul>	s may e threats may be present.	with any of the	threats described
	are looking for sources of searison group, but <i>not</i> the c		that
3.4 Any Study			
<ul> <li>Even with good and p woods of threats.</li> </ul>	oroper use of comparison (	groups, we are not	entirely out of the
3.4.1 Observer Bias			
When we described of bias in how our research	observational measures, warch observers	•	oblem of potential ord measurements.
-	oncern in experiments, eve construct validity impacts th		rous procedures -
• Prevention:			

<ul> <li>Remember to of the experiments groups and put</li> </ul>	your observers, i.e., make them unaware pose prior to the recording.
3.4.2 Demand Characteristics	
<ul> <li>Review: a demand characteristic is when naturally due to understanding then naturally due to understanding then naturally due to understanding then naturally due to understanding them.</li> </ul>	en a participant changes their behavior unure of the experiment
<ul> <li>Prevention:         <ul> <li>Ideally, both participants and observable</li> <li>hypotheses, to prevent possible</li> </ul> </li> </ul>	ervers should be blinded to study goals and from arising.
<ul> <li>Little detail about the conditions sh</li> </ul>	nould be shared until
3.4.3 Placebo Effect	
<ul> <li>Placebo effects occur when the mere         a pronounced positive effect - this is ex</li> </ul>	in a treatment produces tremely common
<ul> <li>Prevention:         <ul> <li>We can use double-blinding and me</li> <li>sees a more pronounced effect the</li> </ul> </li> </ul>	
3.5 Validity in the Face of Many T	nreats
Despite the numerous challenges     and rigorous designed.	discussed above, experiments are still
, , ,	ders for preventing many of the threats dis- be addressed with specific attention paid to e.
4 Null Effects	
4.1 Overview	
• Sometimes, we fin by a p > 0.05 or a 95% CI that contains	d a difference where we expect, as evidenced 0. What gives?!
<ul> <li>This is actually very common in research having been done</li> </ul>	n, and not necessarily indicative of something

However, we should i the statistical	nvestigate whether changes to t of the results	the design would have changed
4.2 Not Enough Diff	ferences Between Group	<b>OS</b>
One root cause of hat between the indepen	_	ufficient evidence of difference 
4.2.1 Weak Manipulation	n	
•	manipulation (i.e., the difference enough to create a difference in	e between the two conditions) is
<ul> <li>Put another way, the insufficient</li> </ul>	e way we	our construct of interest was
4.2.2 Insensitive Measu	res	
We may have the re	verse problem: where our ma variable measure is not sensiti	nipulation is impactful, but our ve enough to detect it.
<ul> <li>Here, we may need to enough to capture ch</li> </ul>	consider using aange in the outcome.	measure that is detailed
4.2.3 Ceiling and Floor	Effect	
• A <b>ceiling effect</b> ha	ppens when most participant end of possible scores on the o	s have a score close to the outcome measure.
• Inversely, a <b>floor</b>	effect is when most scores of possible scores.	s are clustered towards the
These are symptoms means that are close	s of ame together, regardless of condition	easure, and will often result in on.
4.2.4 Value in Manipulat	tion Checks	
		fer some way to assess whether s often some additional variable

A design confound may confuse the results of an experiment in the	di-
rection of the effect of the intervention, thus causing the appearance of null results.	_
4.3 Too Much Within-group Variation	
• It may be that we do have a difference between our groups, but the variance within each condition is so large (i.e., or error) that it becomes difficult to determine.	
This often manifests as large standard errors / wideintervals.	
4.3.1 Measurement Error	
Measurement error naturally occurs in instrument, but our goal is to limit this as much as possible.	
• A Formula:	
Observed = True + Error • The solution:	
- Ideally: Use tools! e.g., recall chapter 5 and our various assessments of measurement reliability and validity	
- Less ideal: Just measurepeople! The inconsistency will balance out naturally (probably)	
4.3.2 Individual Differences	
• As previously stated, research is, in other words, we capture general trends and mean differences. However, salient-enough individual differences can confuse between groups effects.	
<ul> <li>The solution:         <ul> <li>Bigger ! This dilutes the effects of individuals with odd or unusual characteristics within each group</li> <li>Use a within-subject design, then each person's trait controls for themselves!</li> </ul> </li> </ul>	

4.3.3	Situation Noise	
•	This is a much more real type of noise, in which the literal of the experiment may be playing some un-intended role in the outcome variable.	
•	This is why many experiments are done in barren,lab settings.	
4.4	Statistical Power	
•	<b>Power</b> , in research, refers to the likelihood that a significant effect is found, when that difference is real.	
•	Ideally, we employ methods that lend themselves to a high amount of power such as:groups designs	
	<ul> <li>Larger</li> <li>Sensitive measures</li> <li>Low "noise" and high control</li> </ul>	
•	In addition to helping us get significant results, this also aids in thethe study.	of
4.5	Transparency About Null Effects	

•	A misconception: Non-significant effection	cts are	_worth reporting. _
•	Null findings are	to the process of self-corre	cting science