

# Week 11 Lecture - Experimental Confounds

Undergraduate Research Methods in Psychology

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# **1** Chapter Overview

#### **1.1** Learning Objectives

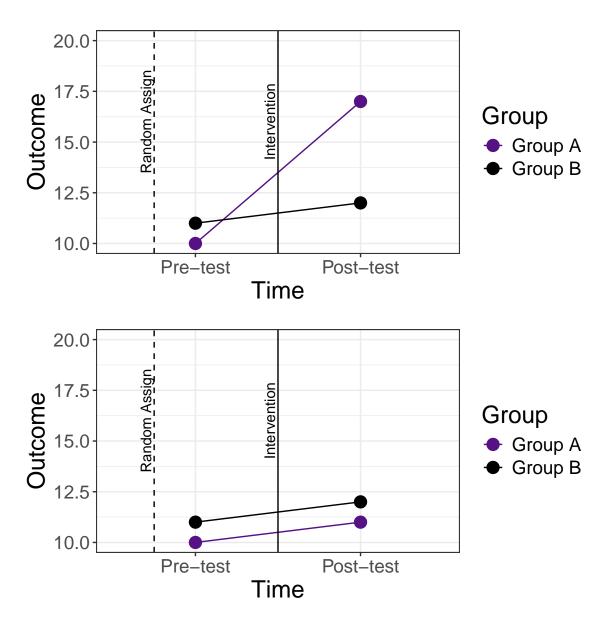
- Interrogate a study and decide whether it rules out twelve potential threats to internal validity.
- Describe how researchers can design studies to prevent internal validity threats.
- Interrogate an experiment with a null result to decide whether the study design obscured an effect or whether there is truly no effect to find.
- Describe how researchers can design studies to minimize possible obscuring factors.
- Describe the importance of comparison groups in ruling out threats to validity
- Explain an example for each type of validity threat

### **1.2** Chapter Overview

ecause of the e cautious of a number of internal va	required to investigate causa alidity threats	al claims, we must
e may have to be especially mindful ur finding are "real". Design flaws m		in order to ensure s.
ailure to design around these threats alidity, and therefore, may hurt our c	_	

# **2** More Internal Validity Threats

## 2.1 Working Example



#### 2.2 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

Discuss: We used two different names for intervention was done on separate groups or the of those types?	•
However, there are many other	to also be concerned with
2.3 For One-group & Pretest/Posttes	st Designs
Some threats are a particular concern in output	ırsubjects designs:
2.3.1 Maturation	
<ul> <li>Certain behaviors may simply this change as being "spontaneous" and u</li> </ul>	by themselves - we may describe nexpected
This is change that is thereof	_ explainable by our intervention or lack-
Discuss: What is an example of a behavinaturally over time, not necessarily due to a s	, ,
because we have a comparison grodespite the maturation.	erhaps there is a maturation effect, but oup, we can see if there is a difference ely demonstrated with a pretest/posttest over

2.3.2 History	
An effect may occur due to some change in our groups.	that creates an unexpected
However, to be a true history threat this must be in a outsized manner on only one of the groups.	, i.e., happening
? Which of the following seems to be the best example o	f a history threat?
A) All members of one condition are older than anot B) The order of our conditions effects how participal C) Some participants in both orders change their be D) Some folks in one condition become extremely il Explanation:	nts respond to an intervention chaviors as they discover the hypothesis
<ul> <li>Prevention:         <ul> <li>Once again, our comparison group saves the day same event, then the effect is for the differences between the groups.</li> <li>However, if the event somehow has a group, the experiment may have to be re-done under the comparison of the differences.</li> </ul> </li> </ul>	and not of major concern impact on one
2.3.3 Regression	
This threat revolves around regression to the extreme scores tend to naturally converge towards the	, where, over time, e central tendency of the data
<ul> <li>Thus, extremely high scores naturally trend downward a low scores naturally trend</li> </ul>	as time passes, and extremely
<ul> <li>Example: Over time, extreme IQ scores tends to conv develop more</li> </ul>	verge to the mean as children
<ul> <li>Prevention:</li> <li>Guess what saves us here?</li> <li>If we see groups equal at the start, and a di</li> </ul>	groups! (Notice a trend?) ifference in trend, we know

that one group did indeed have an effect above and beyond the effects of

2.3.4 Differences Between Maturation,	History, & Regression
Maturation threats deal with havior	and unexplainable change in be-
History threats are due to a mental influence	event or know outside environ-
<ul> <li>Regression threats come form natural the mean or center of a scale.</li> </ul>	lyscores converging on
2.3.5 Attrition	
<ul> <li>Attrition occurs whenever we have so sample (may also be known as "mortal")</li> </ul>	<u> </u>
<ul> <li>If attrition happens among all group concern is low.</li> </ul>	os and of people, then
But, If there is high attrition among of that could	ne of the groups, we have a differential effect
Discuss: Other than death, what are leave the sample?	some other reasons why a participant may
<ul> <li>Prevention:         <ul> <li>It is often impossible to fully pr</li> <li>(because research is</li> </ul> </li> </ul>	revent people from not completing the study !)
<ul> <li>Instead, we may opt to data corresponding to those who</li> </ul>	delete, or completely remove the or dropped out.
2.3.6 Testing	
As we learned with	effects, participants may grow more skilled

if they take an assessment more than once, not due to intervention, but just due to

natural growth.

<ul> <li>Prevention: <ul> <li>Don't use a</li> <li>threats more difficult</li> </ul> </li> <li>Use two different, but equivalent they measure the same thing?</li> </ul>		
2.3.7 Instrumentation		
<ul> <li>This occurs when something about to over time.</li> </ul>	he	instrument changes
<ul> <li>Put another way, our measurement is</li> </ul>	not behaving	<u>!</u>
Discuss: Try describing the different	between a testing and a	n instrumentation
Prevention:		
<ul> <li>We may use only a post-test de over time if it is only used once!</li> </ul>	sign. The	can't change
We may use construct validity sta	tistics (e.g.,	lpha) to assess
validity and reliability of the mea  – We may counterbalance order o	•	forms
2.3.8 Combined Threats		
<ul> <li>Our selection effects may here, or even multiple threats may be</li> </ul>		he threats described
<ul> <li>As a general rule, we are looking for that may affect one comparison group</li> </ul>	_	

? I have found that members of all 3 condit personality types, regardless of the group and should I be concerned?	
A) Unsystematic, and should be concer B) Systematic, and should be concer C) Unsystematic, and should not be D) Systematic, and should not be concer Explanation:	rned concerned
2.4 Any Study	
Even with good and proper use of out of the woods of threats.	groups, we are not entirely
2.4.1 Observer Bias	
<ul> <li>When we described observational mea bias in how our research observers</li> </ul>	sures, we broached the problem of potential and record measurements
This is still a conc procedures - effectively, a part of consti	ern in experiments, even despite the rigorous ruct validity impacts the internal validity
Discuss: Try explaining how the constr bias also contribute to problems in internal	· · · · · · · · · · · · · · · · · · ·
• Prevention:  - Remember to	your observers, i.e., make them unaware
of the experiments groups and pur	pose prior to the recording.

## 2.4.2 Demand Characteristics

• <i>Review:</i> a denoted in naturally due		when a participant cr _ then nature of the e _	nanges their benavior un- xperiment
-	both participants and ol eses, to prevent possible		linded to study goals and _ from arising.
<ul><li>Little de</li></ul>	tail about the conditions	should be shared un	til
2.4.3 Placebo E	ffect		
	cts occur when the mere d positive effect - this is	-	in a treatment produces
_	ndition a sugar pill with n, but I do not tell them v roups?		
A) Sugar p	oill - placebo; medication	ı - treatment	
B) Sugar p	oill - control; medication	- control	
C) Sugar p	oill - treatment; medication	on - treatment	
D) Sugar p	oill - treatment; medication	on - placebo	
Explanation:			
• Prevention:			
	ı use double-blinding ar ees a more pronounced		
group s	ees a more pronounced	enect than the place	ου group
2.5 Validity ir	n the Face of Many	Threats	
• Despite the	numerous challenge and rigorous d		, experiments are still
•	• .	lly be addressed with	g many of the threats dis- specific attention paid to

# 3 Null Effects

-	_		
2 1		rv/I av	ı
3.1	UVE	rview	,

• Sometimes, we denced by a p > $0.05$ or a 95% CI that	find a difference where we expect, as evicontains 0. What gives?!
This is actually very common in research having been done	h, and not necessarily indicative of something 
	er changes to the design would have changed the results
3.2 Not Enough Differences Betw	een Groups
One root cause of having null effects between the independent variable	is having insufficient evidence of difference
3.2.1 Weak Manipulation	
<ul> <li>It is possible that our manipulation (i.e., simply not impactful enough to create</li> </ul>	the difference between the two conditions) is a difference in
<ul> <li>Put another way, the way we</li> <li>insufficient</li> </ul>	our construct of interest was
? Previously, we used a vocab term to de had the intended effect - what was that ter	
A) Manipulation check	
B) Manipulation insurance	
C) Pilot block	
D) Pre-test	
Explanation:	

#### 3.2.2 Insensitive Measures

We may have the reverse problem: where our manipulation is impactful, but our variable measure is not sensitive enough to detect it.
Here, we may need to consider using a measure that is detailed enough to capture change in the outcome.
3.2.3 Ceiling and Floor Effect
• A <b>ceiling effect</b> happens when most participants have a score close to the end of possible scores on the outcome measure.
Inversely, a floor effect is when most scores are clustered towards the of possible scores.
Discuss: Try to come up of an example of when you might run into a ceiling effect
These are symptoms of a measure, and will often result in means that are close together, regardless of condition.
3.2.4 Value in Manipulation Checks
We alluded to <b>manipulation checks</b> before, which offer some way to assess whether a manipulation truly caused any sort of change. This is often some additional variable we alongside the outcome.
3.2.5 Reverse Design Confound
A design confound may confuse the results of an experiment in the
direction of the effect of the intervention, thus causing the appearance of null results.

# 3.3 Too Much Within-group Variation

to determine.	groups, but the variance within or error) that it becomes difficult
to determine.	
This often manifests as large standard errors / wide	intervals
? What validity are standard errors and confidence into	ervals associated with?
A) External validity	
B) Internal validity	
C) Construct validity	
D) Statistical validity	
Explanation:	
Measurement error naturally occurs in is to limit this as much as possible.	instrument, but our goal
• A Formula:	
Observed = True + Err	ror
$Observed = True + Err \\ \bullet \ \textit{The solution:}$	recall chapter 5 and our various
Observed = True + Err • The solution: - Ideally: Use tools! e.g., assessments of measurement reliability and value - Less ideal: Just measure	recall chapter 5 and our various alidity

	<ul><li>Bigger unusual characteristics withi</li><li>Use a within-subject design,</li></ul>	• .	
3.3.3	Situation Noise		
	nis is a much more real type of e experiment may be playing so		of tcome variable.
• Th	nis is why many experiments are	done in barren,	lab settings.
3.4 S	Statistical Power		
	<b>ower</b> , in research, refers to the und, when that difference is rea		significant effect is
• Id	eally, we employ methods that letgroup: - Larger - Sensitive measures - Low "noise" and high control	s designs -	nt of power such as:
In addition to helping us get significant results, this also aids in the     of the study.			
3.5 Transparency About Null Effects			
• A	misconception: Non-significant	effects are	worth reporting.
• N	ull findings are	to the process of self-cor	recting science
Discuss: Explain what purpose null results serve in the Theory-Hypothesis-Data framework			