

# Week 13 Lecture - Quasi-experiments & Small-n Designs

Undergraduate Research Methods in Psychology

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# Chapter Overview

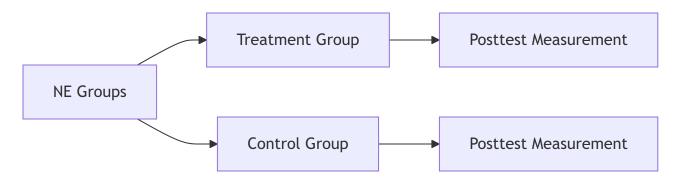
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<del>-</del>	usal effect is cool!  methods, which we calldesigns for when we have
<ul><li>2 Quasi-Experiments</li><li>2.1 Quasi-experimental Variables</li></ul>	
<ul> <li>Unlike with exp</li> <li>experiments do not have full control of</li> <li>E.g., assignment to different class istrators, not researchers</li> <li>These scenarios involve some some some some some some some som</li></ul>	eriments, like we've discussed so far, quasi- f the IV srooms or different school programs by admin- ort of "conditions" but those conditions are not assignment, like we would have in experiments
	e examples of times that you might run into

# 2.2 Types

#### 2.2.1 Non-equivalent Control Group Posttest-only Design

- Similar to our previous groups posttest-only design, this design will have two or more comparison group, both measured on an outcome
  - In practice, these will often be analyzed with the same tests, and look very similar in structure
- However, they are still only tested \_\_\_\_\_ the "intervention" occurs, not before

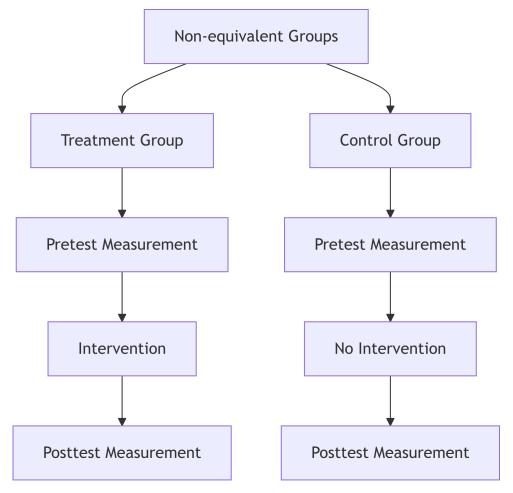


- In the "Quasi" version, the researchers have the ability to randomly assign who goes in the experimental or control groups
- *Example:* Researchers look at folks treated with two different medications for high blood pressure after admission to the hospital, in a retrospective study.

Discuss: Where does the non-equivalent part of the design name come from; what makes the normal experimental posttest equivalent groups?

#### 2.2.2 Non-equivalent Control Group Pretest/Posttest Design

• Largely an \_\_\_\_\_ of above, this follow the same procedures as the previous design, but now includes measurements before and after the intervention.

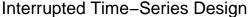


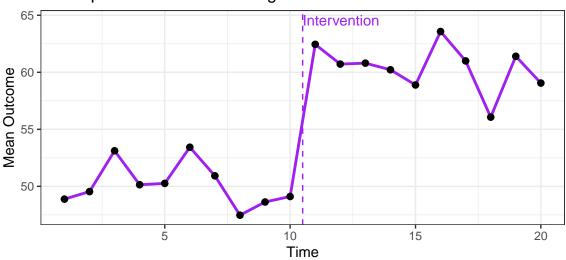
- *Example:* The same medical researchers from the previous example now compare blood pressures at outpatient visits prior to admission, and then changes in blood pressure after treatment with two different medication during admission
- ? What previous true experimental design is this an analog to?
  - A) Pretest/Posttest
  - B) Longitudinal
  - C) Multiple Regression
  - D) Mixed Factorial

**Explanation:** 

#### 2.2.3 Interrupted Time-series Design

- This is when we are \_\_\_\_\_ some variable for a period of time, and then it's measurement is "interrupted" by some clear event
  - We can then compare folks on the \_\_\_\_\_ and after of the measured variable after the event
- This type of design can appear to be remarkably similar to a multivariate correlational design - but with the added confound of some meaningful event occurring
- Example: I am tracking changes in optimism weekly in teenagers about the current political climate, and then a national election happens and I continue to gather weekly info

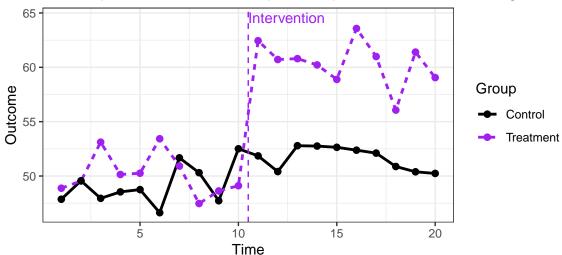




## 2.2.4 Non-equivalent Control Group Interrupted Time Series Design

- A \_\_\_\_\_ of the time series and non-equivalent control group designs, where we have both comparison groups and also a historical event that interrupts some variable measured over time.
- This behaves similar to a \_\_\_\_\_ design because we place emphasis on both our separate groups (between) and how they differ from before and after the event (within)





• Example: Same as the last example, but now I track how men and women differ in optimism levels at multiple points before and after the election

Discuss: Review - what part of factorial designs tends to be the most interesting (hint: it looks like a crossover between lines on a graph)

# 2.3 Internal Validity

- Just like with \_\_\_\_\_\_ experiments, these designs are similarly affected by threats to internal validity

   Recall the many threats we've talked about already for validity

   We can still make \_\_\_\_\_ to control for these possibilities, but we should be cautious on how the "quasi-" part of these studies are cause for concern as we do in true experiments!
- Thankfully, the \_\_\_\_\_ for many of these are similar to their non-quasi counterparts!

	Selection Effects  Recall that a selection effect occurs if there is group receiving one condition of the IV.	variability in one
•	This can be especially lack of random assignment - which is largely	to account for, because of the general how we prevent these effects normally
?	Which of the following was not an internal val	idity threat?
Ex	A) Regression to the Mean B) Attrition C) Maturation Effects D) Large sample size planation:	
•	Solution:	
	<ul> <li>Carefully monitor and consider</li> <li>groups</li> </ul>	differences between
	<ul> <li>Use a pretest/posttest design to see diff</li> </ul>	erent points and
	trends over time	groups to "agualiza" pro
	<ul> <li>More complex and difficult:</li> <li>existing differences</li> </ul>	groups to "equalize" pre-
	If we clearly measured the variable the study a study as a second control of the second control of the study as a second control of the se	
2.3.2	2 Design Confound	
•	Similar to above, but this is when there is so the time as the change	
	<ul> <li>Because we don't necessarily have solic independent variable, it is possible there the same time</li> </ul>	•
_	Example: An earthquake happens among th	سمؤ بموانين مممور مرموا الممطل المطلوب بمريم

- Example: An earthquake happens among the group that I had been measuring for happiness, but on top of this, there was severe change in economic status of the country at the same time - so where did a change come from?
- There isn't a full proof solution to this, sometimes we just may have to recognize a that the change is due to a multi-faceted change

<b>4</b>	Discuss: What part of a published article are limitations most often discussed?
2.3.3	Maturation Effects
	Like as with the "normal" maturation effects, we can account for this by observing comparison groups and using a pretest/posttest design.
2.3.4	History Effect
	First thing to consider is whether a history effect cause systematic variability in only one
•	Solution:  - Still using comparison groups and pretest/posttest!
2.3.5	Regression to the Mean
•	Surprise, surprise, this is still the same as discussed before
•	Solution:  – Still using comparison groups and pretest/posttest!
2.3.6	Attrition
	Be mindful to check for attrition effects across variables and IV conditions.
2.3.7	Testing & Instrumentation
	Largely, just ensure construct validity and use parallel forms to prevent practice effects.

Discuss: Describe, in your own words, what the diffe Instrumentation threats is	rence between Testing and
2.3.8 Observer Bias, Demand Characteristics, & Place	ebo Effects
Observer bias is only present if we use	measures
characteristics will be minimized unaware of what "condition" they are in	d if participants are blinded and
Placebo are only a concern whe receiving an inert treatment, and can be nullified with	n we have a comparison group a control group.
2.4 Priorities of Validity  2.4.1 Real-world Opportunity	
Sometimes, societal change presents an interesting wouldn't otherwise be possible on such a	question for researchers, that
Example: Something like COVID had profound effects we can't easily control things like lockdowns - it is still	
<ul> <li>Also, sometimes studies become quasi-experimental by changes during the study.</li> </ul>	y virtue of
Discuss: What are some other events that would question?	spur a quasi-experimental

- Magnitude (effect size)

- Precision (confidence intervals)

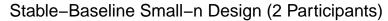
2.4.2	External	Validity
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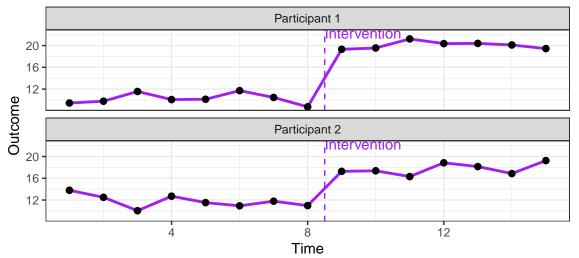
<ul> <li>In some ways, these types of studies are more participants in a more natural environment, en</li> </ul>	
But still watch out for sampling	<u>!</u>
2.4.3 Ethics	
<ul> <li>Like with Real-world Opportunity, quasi-exper curring groups that wouldn't be ethical to</li> </ul>	iments may be done on naturally oc-
Example: Those who suffer an adverse react who do not	
? Review - Which of the follow was one of the ethic Report?	cal principals present in the Belmont
A) Ethical Review	
B) Justice	
C) IRBs	
D) Debriefing	
Explanation:	
2.4.4 Construct and Statistical Validity	
<ul> <li>Just like with previous studies, our construct variables are captured in</li> </ul>	
Statistical validity is:	

2 5	With	Corre	lational	Studies
<b>Z</b> .J	VVILII	CUIT	ialiviiai	Junica

The primary studies is intentions and events	between the correlational and	quasi-experimental			
<ul> <li>Quasi-experiments usually are lo</li> </ul>	ooking at a specified	events or			
external events, whereas correlational studies deal more with >just casting a wider net and measuring internal naturally occurring >phenomena.					
<ul> <li>Importantly, some studies may be</li> </ul>	oe okay to describe	ways			
2.6 Quasi-independent vs. Participant Variables					
Quasi-independent	are primarily those that	change over a large			
portion of society or people	are primarily those that	onange over a large			
<ul> <li>Participant Variable are internal,</li> </ul>	, charactei	ristics, such as age,			
gender, race, diagnosis, etc.					
<ul> <li>Quasi-experiments do not result from differences in participant variables, that is saved more for correlational designs that we already discussed.</li> </ul>					
3 Small-n 3.1 Overview					
<ul> <li>Small-n designs are unique in t Sometimes, it is just one persor</li> </ul>		·			
? What do you guess is a common	n name for a intense study done	on one person?			
A) Independent Study					
B) Intensive Study					
C) Case Study					
D) Briefer					
Explanation:					
·					

This is often due to often	our group of interest to each participant	•	I, or an extreme time
The     ualized impact, which     quantitative research	h is a departure fron	dies are often more c n the traditional proba	concerned on individ- abilistic goals of most
3.2 Core Character	istics		
• Each person is treate	d as an individual, ra	ther than	with others.
• Data is not	(i.e., tu	rned into a mean or m	nedian)
- No other	stati	stics like standard dev	viation either!
Discuss: Review - If why is that?	we can't do descriptiv	ve statistics, we can't c	lo inferential either,
• Designs are used to o	closely monitor timinç	g and	to interventions.
Often used in therape	eutic or care settings	5	
33 Tynes			
<ul><li>3.3 Types</li><li>Small-n designs all</li></ul>	charo a rolativolv	emall cample cize	but have different
- Smail-II designs an	Share a relatively -	Small Sample Size,	, but have unlerent
3.3.1 Stable-baseline			
This is when a person time to observe an un-	•	neld at a some outcome varial	for sufficient ble.
<ul> <li>This baseline period and more measurem</li> </ul>	•	vith some	or change,





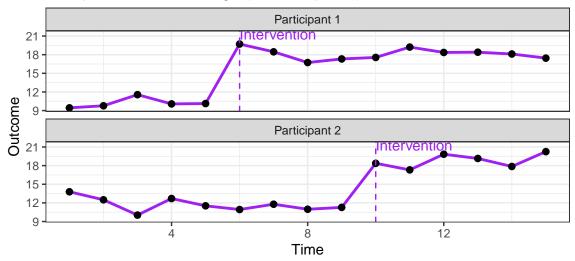
• Visually, this is almost as if you did a pre-test/post-test \_\_\_\_\_\_, but with just one person!

*Example:* I monitor a person's depressive symptoms for several months, and then start a new medication and see change afterward

#### 3.3.2 Multiple-baseline Design

- This design requires people, and necessitates staggering the timing of the intervention across the participants, to see if the timing alone is explanatory in the change.
- It also helps in observing whether multiple participants see the same \_\_\_\_\_
   of behavior change, regardless of timing

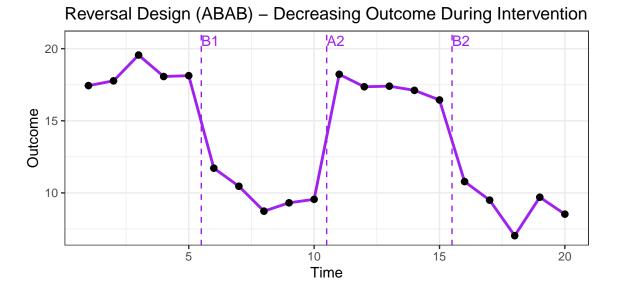




Discuss: If we have multiple (albeit few) people, why not switch to one of the other experiment types we've talked about?

### 3.3.3 Reversal Design

- This is when a naturally undesirable is allowed to occur at baseline, and then a therapy is applied to reduce it.
- Then, after sometime, the treatment is removed, and the \_\_\_\_\_ of the behavior is analyzed.



#### 3.3.4 Single-n

- This is a general category term that captures any study that looks at only one person over usually a period of time.
  - It is common that multiple measurements and interventions for this person may be gathered the study.

# 3.4 Balancing Priorities in Small-n

- These studies are naturally very limited in their ability to \_\_\_\_\_ to other situations and people, due to the uniqueness of the person under study.
- However, they tend to be useful in examining and describing \_\_\_\_\_ or unique cases that cannot be replicated - and some implication may inform directions in future research.

# 3.5 Disadvantages of Small-n

- · · · · · · · · · · · · · · · · · · ·		
• Without comparison gi	roups, we often open ourselves u <sub>l</sub>	p to numerous internal validity
– E.g.,	threats, regression th	reats, etc.
External validity will tend to be naturallyspecific to individual tendencies.		as the cases are so
•	an hardly be considered	of many people

Discuss: In the case of a single from personal experience?	person being studied,	, what separates this research	
			_
3.6 Assessing Validity in Si	nall-n		
<ul> <li>Internal validity can be reason</li> </ul>	ably	, especially in the case o	of
multiple baseline and reversal (	•		
<ul> <li>Like with any design, the was</li> </ul>	central question to it for possible confour	nternal validity is whether ther nds.	е
• External validity is relatively wea	k, but may be	more with furthe	er
larger studies.			
<ul><li>And not all</li></ul>	need to gene		
– E.g., a clinician working w	-	•	
Construct validity is assessed justine in the construct validity is assessed justine.	st like any study - with	the use of	
bias tools and observations.	ctatictic	os for tools, as well as author	_
<ul> <li>I.e., look at the explanations and rationale</li> </ul>		cs for tools, as well as author	5
Statistical validity tends to ofter		cal than truly statistical (becaus	<u>:</u> _
<u> </u>	istics require large g	•	
<ul><li>E.g., our trusty friend,</li></ul>	ple	ots!	
	·		