

Week 9 Lecture - Multivariate Correlation

Undergraduate Research Methods in Psychology

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Objectives and Overview 1

Learning Objective(s) 1.1

- State why simple bivariate correlations are not sufficient for establishing causation.
- Explain how longitudinal correlational designs can establish temporal precedence.
- Explain how multiple-regression analyses can rule out some (but not all) third variables.
- Describe the value of pattern and parsimony, in which a variety of research results support a single, parsimonious causal theory.
- Explain the function of a mediating variable.
- Understand the relative strengths and weaknesses of a multivariate design, compared to bivariate designs
- Be able to describe interpreting longitudinal and multiple regression designs, and interpret statistical values from these designs

1.2 Chapter Overview

 Covariance: are tent pattern?

•	$\begin{array}{c} \longrightarrow \text{ multiple (n} \\ \hline \text{- We will still be using a lot of co} \end{array}$	nore than two) variables rrelation-like techniques	
•	Example: In a iety that I am looking at (two variable depression, anxiety, well-being, and	design, I might have just depression les). In a multivariate version, may I academic success.	
•	This correlational research is <i>still</i> con – We still don't have manipulation	· · · -	variables
•	•	ssed today will get us closer to a caung the necessary criteria, and still in	
2	Review of the Casual C	riteria	
2.1	Overview		

(or more) variables varying in some consis-

• Temporal Precedence: time?	does one variable come	another in
• Vale explained by a third varia	•	ween our variables not better
? Which of these is already	met by a bivariate design?	
A) Covariance B) Temporal precedence C) Internal validity D) None of them Explanation:	ce	
3 Longitudinal Des 3.1 Overview	signs and Tempora	l Precedence
 Longitudinal Research: Research conducted an extended period Normally follows the measures at the ma 	of time) group (points (usually over of people taking same multiple
	s rating of their own physical	and mental health health over
	of this comes from the 2 or meime points	ore measurements happening
- In essence, the sam because they fall at	e measures are treated as _ different time points	variables
 With this design, comes each type separately in a 	· · · · · · · · · · · · · · · · · · ·	, we will discuss

?

3.2 Cross-sectional Correlation



- Are the variables at the same time point correlating with one
- The refer to the separate measures
- ? Which of the following r values suggest at least weak relationship?

A) r = .007

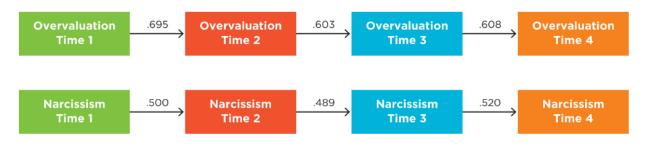
B) r = .070

C) r = .138

D) r = .099

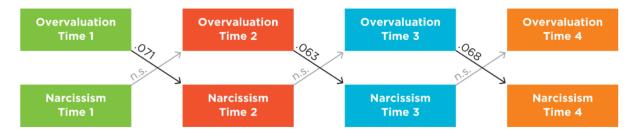
Explanation:

3.3 Autocorrelations



- Are the variables correlated with at the other time points?
- "Auto-" as a root prefix means "self", so this is a measure correlated with itself

3.4 Cross-lag Correlation



- Are the variables correlated with each other time points?
- This is sufficient for establishing
- This is also the _____ outcome we are looking at for longitudinal designs
- The "cross" here should remind you of the diagonal direction of the arrow in this diagram

Discuss: Normally these are interval/ratio variables, but what statistic could we use if they are ordinal?

3.5 Longitudinal and Causation Criteria

- ? What is the one causal criteria we are still missing in a longitudinal design?
 - A) Internal validity
 - B) Covariance
 - C) Temporal precedence
 - D) All of them are met

Explanation:

an entire month.

•	ergy level may be affecting the relationship lf-worth, I should measure that as well.
3.6 Why Not Experiment	
You can't always easily thing	variables that may be causing some-
Discuss: What are examples of variable reasons?	s that are difficult to manipulate for practical
You can't always	manipulate a variable
Discuss: What are examples of variable reasons?	es that are difficult to manipulate for ethical
longitudinal designs.	study that would be unethical to over n combine those experimental designs with eople for one night, but can't do it ethically for

4 Multiple Regression and Third Variables

4.1	Overview			
•	Multiple	regr	ession is when we predict one contin	uous variable
	by way of mult	iple	variables	
	Stats side	bar: some books ter	mporarily refer to this as "multivariate is a separate, more complicated tec	•
•	•	e can predict with $_$	or continuous v	/ariables, but
	 Stats side 		itinuous nultiple regression <i>can</i> be binary/ca nplicated and won't be our focus in t	-
4.2	Criterion a	nd Predictor Va	riables	
•	When we use	multiple regression, the "de _l	we have: pendent" variable we are trying to pi	redict
•	rion (duh)		at we are using toiables $ ightarrow$ single criterion variable	the crite-
4.3	More Than	Two Variables		
•	The benefit to	this is we can	control for the eff	fects of other
	variables and most		the effects of the variables we ca	are about the



Exposure to sexual TV content

4.4 Statistics for Third Variables

When we talk about <i>controlling</i> for some about "holding it still" or	variable in a model, we are talking the effect it has on the criterion or
That way, we can see the individual on the criterion	of each predictor variable
 This is not the same as making a variable a this is purely a statistical procedure 	in the design itself

4.5 Beta coefficients

• Beta, represented as β , is a _____ effect that one unit on the predictor variable changes on the outcome variable

 For example, if our age / year of age, 		ancy risk - that means that for every 1 y risks raises by 1.25.
• A beta further from 0 sig	nifies a	effect, and we can look for
statistical	in eta just like	in other stats.
Discuss: If I have a beta tation of the relationship betw	<u>'</u>	X for criterion Y, provide an interpreles.

4.6 Interpretation Example

Multiple-Regression Results from a Study Predicting Pregnancy from Sexual Content on TV and Age

CRITERION (DEPENDENT) VARIABLE: PREGNANCY RISK	ВЕТА	95% CI FOR BETA	Statistical Significance
Predictor (independent) variables:			
Exposure to sex on TV	0.25	[.14, .36]	*
Age	0.33	[.20, .46]	*

Note: Data are fabricated, based on imagined results if the researchers had used only two predictor variables.

4.7 More Predictors, The Merrier?

There is no natural limit to however many pre	edictors to put into a model, with some
caveats:	
 A good rule of thumb is that sample sh 	ould be at least
times the number of predictor variables	
 Too many predictors can 	effects if the predictors correlate
with one another!	

^{*}p < .05, meaning the result is statistically significant and the 95% CI does not include zero.

	Discuss: AI is sometimes descr n model - why do you think this	ribed as an especially complicated multiplies?	le regres-
4.8	In Popular Media		
•		, "Considering" are all phrases that often lel. Be on the lookout for these.	indicate use
4.9	Still Not Causal		
•		and multiple regression studies are	
	at statistically control of third va	ariable and establishing temporal preced	ence
•	They don't quite reach the	gold standard of expe	riments
5	Pattern and Parsimo	ny	
5.1	Overview		
•	Sometimes the appears to be sufficient for esta	of evidence from substantial correla ablishing causality is it?	ation studies
5.2	Meaning		
•	<u> </u>	nciple that says we should seek the simpl of a phenomenon or relationship	lest possible
•	The "pattern" we are speaking of from numerous studies	of here is the pattern of	coming

? What type of source is going to give a nice quantitative overview of the weight of evidence in an area?
A) Original empirical journal article B) Meta-analysis journal article C) Literature review journal article
D) Scientific journalism

Explanation:

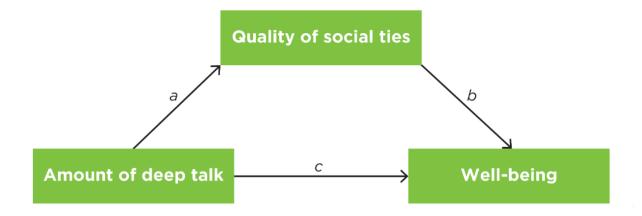
5.3 In Popular Media

 Popular media tends to like big 	papers that seem to point out
a prominent pattern in research	
However, remember the	associated with putting too much trust
in scientific journalism	

6 Mediation

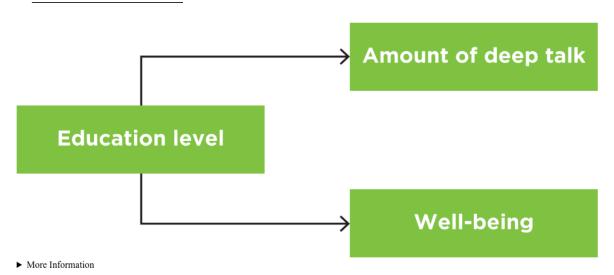
6.1 Overview

• Mediation: a claim th	at one variable's relationship with another is bette
by a	another variable, i.e., some mechanism through which ar
effect occurs - the "why"	
• Technically this is a designs.	claim, but is often hinted at via correlationa



6.2 Compared to "Third Variable"

• A "third variable" more so refers to a variable unrelated to a mechanism that just to be related to both variables of interest



6.3 Compared to "Moderators"

- A _____ is best described as a variable that's state changes the relationship between two others
- E.g., The link between depression and anxiety are stronger when trauma is high

Discuss: How common do you suspect mediators and moderators are in psychological phenomena?		
7 Analysis with the Four Validities		
7.1 Overview		
• How do we investigate these designs? Largely the same as the bivariate designs!		
7.1.1 External Validity		
 How was the sample ? To a lesser extent, how many people 		
are in the sample, and what are their demographics?		
7.1.2 Statistical Validity		
What is our (effect size)?		
What is our precision (intervals)?		
Are our statistics?		
Do these results in other studies?		
Do we have a restriction of or outliers?		
? Which of the following scenarios seems most indicative of poor precision?		
A) We have an r-squared of 0.25		
B) Our p-value is 0.10		
C) Our effect isn't shown in other studies D) Our confidence intervals for r are [0.20, 0.80]		

Explanation:	
7.1.3 Construct Validity	
Are our measures consistent and	?
Are our measures accurate and	to our intended construct?