

# **Week 5 Lecture - Measurement**

Undergraduate Research Methods in Psychology

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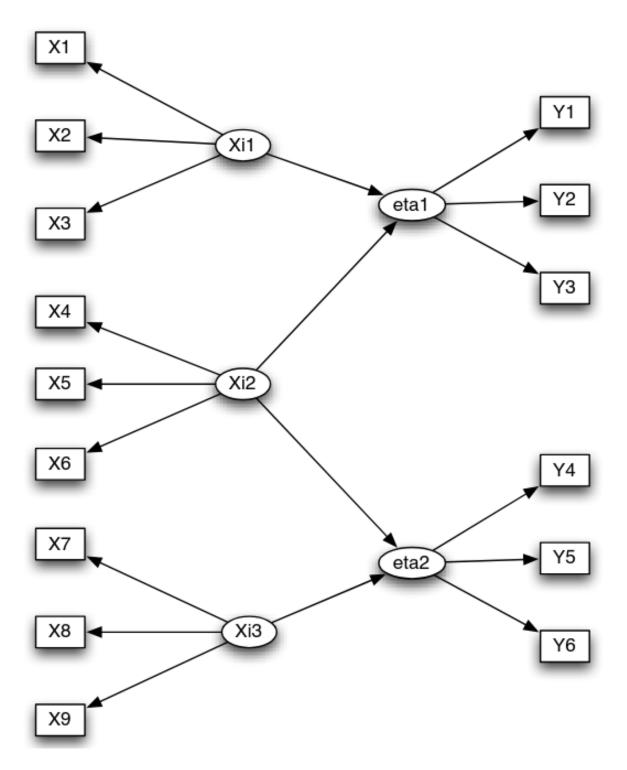
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Ta	able of Contents	
1	Chapter Overview21.1 Introduction to Measurement	
2	How to Measure Something?         2.1       Overview	
3	Reliability (Consistency)63.1 Overview63.2 Three Types of Reliability63.3 Scatterplot Visualization63.3.1 Perfect Positive Relationship73.3.2 Imperfect Positive Relationship73.4 Correlation Coefficient73.4.1 Perfect Positive Relationship83.4.2 Imperfect Positive Relationship83.4.3 Internal Reliability83.5 Reading About Reliability9	
4	Measurement Validity (Accuracy)94.1 Overview94.2 Measurement Validity94.3 Face & Content Validity104.4 Criterion Validity104.4.1 Correlative Methods (for Continuous Behavior)104.4.2 Known Groups Methods (for Categorical Behavior Groups)114.5 Convergent & Divergent Validity114.5.1 Convergence with Related Measures114.5.2 Divergence with Unrelated Measures114.6 Relationship Between Reliability and Validity12	

# 1 Chapter Overview

## 1.1 Introduction to Measurement

Valid and reliable     titative research - without it, it is on the property of the prope	is an essential part of any good quandifficult to test for differences, associations,
<ul> <li>We must be systemic, rigorous, and report (through our writing) thoroughly and experiences.</li> </ul>	in our measurement, and on the methods we use to capture phenomena
	e ways, more difficult to measure than phenom- . For example:
<ul> <li>A chemist is able to</li> <li>cylinders, etc.</li> <li>Physicists can measure weight, n and scales</li> </ul>	parts of solutions with pH, graduated
<ul> <li>Biologists can measure</li> </ul>	of animals or number of times a
certain trait appears in a creatur  – In psychology, we cannot  are internal to people, and even	measure cognitive traits that in behaviors things may be complicated
Remember that for ables from latent/conceptual/construction.	validity: we must make operational variet variables, and we must do this well!
1.2 Visual Representation	
• Rectangles -> Observed Variables (C	Our Measurements)
• Ellipses -> Latent Variables (Our Con	nstructs)
We want to use strategies that	the link between the two



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# 2 How to Measure Something?

**2.3** Three Types of Measure

2.1	. Overv			
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<b>_</b>	OVEI	•		v v

• There	-	construct validit	y of a study.
• There	e are also different mediums fo	or measurement	:
-	-report		
_			
-			
2.2 Cor	nstructs vs. Observed V	ariables (Ag	ain)
•	-	dy, we must co	: ome up with some conceptual I description of a construct. This
- !	usually involves having a reas	sonable knowled	lge of and
	theoretical work in a certain to Then, we must link that conce that fully	•	o an operational measure or tool
	Note: different measures for the	-	may have very dif-
	ferent underlying conceptual c		erstand the
	that your tool makes before yon ple: take the concept of "	ou use it.	" - what even is intelligence?
	Depends on whom we ask: \	Weschler savs	from Binet
: _ '	says different from	_	erent from emotional intelligence e
- 1	Do we take into account age?	)	level? Race? Socioeco-
	nomic background? This is why having a clear liter readers understand a		d background for a tool can help description which goes into the
I	measure of choice		

• All three types have drawbacks and biases which will be discussed more in week 6

2.3.1	Self-Report
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<ul> <li>This is a ing some amount of introsp</li> </ul>		rson it is measuring, often requir-
<ul> <li>Example: Ever been to a c We would call that self-rep</li> </ul>		to fill out a bunch of paperwork?
<ul> <li>This can be either through a naire</li> </ul>	a paper form or through a	a question-
		called a teacher, friend) providing their
2.3.2 Observational		
This is derived from a third	party	a person's behavior/actions
and manner the behavior occur	•	tain behavior occurs or in what
<ul> <li>What I do every day in clir person with some task or sti on a test</li> </ul>		ervational measure: I present a esponse or
2.3.3 Physiological		
This is some sort of measure tends to be much more of a		characteristics of a person, nt that than other two described
above		tions with the types of measures anxiety may also show a specific
pattern of	in the brain du	
<ul><li>Examples:</li><li>Brain scans: CT, MRI</li><li>Facial movement: EN</li></ul>		
<ul> <li>Ideally, we may choose to combination of two of the construct, and they should</li> </ul>	m, to provide multiple o	of measures or some perationalizations for the same (correlated) with one another.

# 3 Reliability (Consistency)

and to the

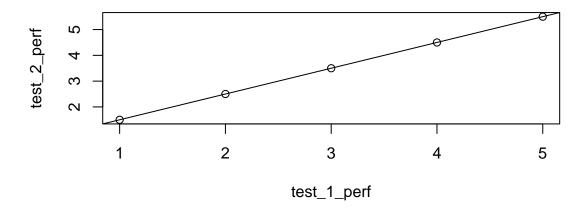
other generally does as well.

3.1 Overview
• is all about how <i>consistent</i> a certain scale or measurement i across different raters, times, and contexts.
We want a measure to be reliable, otherwise, we have a tool that may very well te us a answer every time we take a measurement!
3.2 Three Types of Reliability
<ul> <li>There are generally 3 types of reliability:</li> <li>Test-retest: Between different points</li> </ul>
<ul> <li>Interrater: Between different observers/raters - how often are they rating some thing the</li> </ul>
- Internal: Between items on the measure - how well are related questions regarding the same construct co-varying with one another?
3.3 Scatterplot Visualization
• We may logically approach questions of reliability similar to any other claim of
– In test-retest, we claim that the measure scores at two different times are
<ul> <li>In, we claim that the measure scores, as recorded be each observer, are associated with one another</li> <li>In internal, we claim that two or more on the same measure, for the same construct, covary with one another</li> </ul>
Graphically, we may use a when we have two sets of continuous data, e.g., two sets of scores of any of the above 3 types
The more the points sit close to the line of best fit (which is usually an OLS linear regression line), the stronger the between the two measures.
<ul> <li>In the case of reliability, we would usually like to see a positive relationship between</li> </ul>

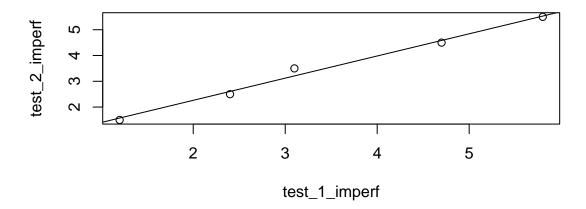
the two sets of data, which would be represented by a line of best fit traveling up

. Put another way, as one set of data increases, the

#### 3.3.1 Perfect Positive Relationship



### 3.3.2 Imperfect Positive Relationship



#### 3.4 Correlation Coefficient

• To mathematically the direction and strength of relationship between two variables we may use r. The type we will be talking about now is technically called Pearson's product-moment correlation coefficient r.

<ul> <li>r will always be between -1 and 1, with 0 representing correlation or relationship and -1/1 representing a perfectly strong relationship between the two. In practice, you will never get exactly 0 or -1/1, but likely some number in</li> </ul>
$r = \frac{\sum{(x_i - \bar{x})(y_i - \bar{y})}}{\sqrt{\sum{(x_i - \bar{x})^2\sum{(y_i - \bar{y})^2}}}}$ • This correlation coefficient is technically only applicable to continuous data, but is to ordinal data by using Spearman's $\rho$ (rho) or Kendall's $\tau - E$ (tau-B).
3.4.1 Perfect Positive Relationship
[1] 1
3.4.2 Imperfect Positive Relationship
[1] 0.9944883
• The $\phantom{aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa$
• Another popular statistic for inter-rater reliability is called Cohen's $\kappa$ (kappa), but that is only when raters are grouping/classifying objects or people It's interpretation is the same as $r$ .
3.4.3 Internal Reliability
$\bullet$ The simplest way to arrange this is to use a correlation of all the items of a measure, and calculate the $r$ between each two items.
<ul> <li>In such a table, we are looking to make sure construct-related items are positive highly and that theoretical unrelated items are negative of weakly correlated.</li> </ul>
<ul> <li>You may also calculate an inter-item correlation which is just an average correlation across the entire matrix (only recommended if all items should be related)</li> <li>Side note: I wouldn't recommend using the acronym AIC for this - most statisticians use that more often for the Akaike information criterion, used often in</li> </ul>
regression  - We want between 0.15 and 0.50 for this to be "reliable"

•	Finally, you can take age inter-item correlation and number of	_'s $\alpha$ (alpha) which is taken items on a scale.	n from the aver-
	<ul> <li>This is probably the "preferred" stati</li> <li>Above 0.80 is good for self-reports,</li> </ul>	stic for	reliability ossible
3.5	Reading About Reliability		
•	The most important things to know are to reliability and a basic how to calculate these values by hand (for the conceptual definition of each.	of each one. You do no	
4	Measurement Validity (Acc	uracy)	
4.1	Overview		
•	This is where our about the 4 validities for investigating cluterms "claim validity" and "measurement"		ty I will use the
•	Essentially, validity struct validity, alongside measurement reestablishing construct validity.	y is the second major comeliability. They both are inc	•
•	A "good" measure (i.e., one with good the following mea	construct validity) will ha surement validities, usuall	
4.2	Measurement Validity		
•	Validity is all concerned with how we tool. It is multiface	ell we represent the cor eted and often quite compl	
	the measurement validity of any giver studies.	n tool has to be well-esta	blished across
•	But remember, just like any claim or evice pletely and flawlessly valid - rather the against its validity.	•	nething as com- idence is for or

# 4.3 Face & Content Validity

•	seems like a certain measurement captures the c	alued when a measurement scale
•	<b>Face validity</b> is largely just an assessment of "work?" - may be by the gene	well, does it seem like this would eral public or by experts in a domain
•	<b>Content validity</b> asks whether it appears a meas nents of a theoretical construct. This is usually domain experts which have a strong knowledge construct.	assessed by
•	For a more empirical, albeit subjective approach, sexperts, calculating $r$ or $\kappa$ for a measurement or across the experts. This could be taken as evident on a tool, but it strays clost the expertise of the judges). For more than 2 judges statistics that are applicable to multivariate G-study, D-study, etc.)	the individual items of a measure ence of somewhat decent expert se to an appeal to authority (i.e., dges we would use extensions of
4.4	Criterion Validity	
•	Criterion validity focuses on whether a measurement with behaviors that are also said to be represent selected are also a subjectivalidity can help establish that a measure is related associate with a trait.	ative of the construct. Now those ive choice - but this measurement
4.4.1	Correlative Methods (for Continuous Behavi	or)
•	Once again, we are able to usethose previous discussed.	methods and scatterplots, like
	<ul> <li>In the scatterplot, we would place the behavior on one axis and the measure on the A strong, positive relationship between the two</li> </ul>	
•	of good criterion validity  For example, consider we are developing a collate children where a parent reports how often a child er we sit in a classroom with the child and measure the	ngages in disruptive behavior. Now,

they engage in disruptive behavior. A higher score on the measure should be associated with a higher number of occurrences of disruptive behavior.

#### 4.4.2 Known Groups Methods (for Categorical Behavior Groups)

- We can also assess whether a measure is able to discern differences between some known-groups by some standard.
  - This means that, prior to testing the new measure, we must have some "source of truth" for whether a person belongs to a certain group.
- Example: We have two groups of people, those diagnosed with schizophrenia and those not diagnosed with schizophrenia. We have a continuous measure designed to detect psychotic disorders. Are those individuals with schizophrenia and those without scoring different on this measure?
- In this method, we could use between-groups \_\_\_\_\_\_, such as t-tests and ANOVA to decide whether scores on the measure are significantly different between the known-group members.
  - In modern research, you are more likely to see techniques like logistic regression, receiver operating characteristic (ROC), and area-under-the-curve (AUC) analysis as these methods are able to detect appropriate "cut-off" scores for the measure that discriminate between the two groups with ideal sensitivity (detection of true positives) and specificity (detection of true negatives)

### 4.5 Convergent & Divergent Validity

- We can also determine how a new measure relates to existing measures for the same/different

   Ideally, we want a new measure to strongly correlate with measures for the same construct (convergence) and have little to no relationship with measures for other constructs (divergence).
- Just like the other measurement validities, correlation is our analysis of choice for these

#### **4.5.1** Convergence with Related Measures

• For a valid measure, we want *high convergence* (i.e., strong positive correlation) with tools that are meant to measure the same construct.

#### **4.5.2** Divergence with Unrelated Measures

• For a valid measure, we want high divergence (i.e., no/negative correlation) with

tools that are meant to measure a different construct.

### 4.6 Relationship Between Reliability and Validity

- Measurement validity and reliability are **not** \_\_\_\_\_\_ terms, though they are related.
  - Reliability is a core necessity for a tool to be considered valid, but just because something is reliable, does not make it valid.
  - "A valid tool is reliable"
  - "A reliable tool is not necessarily valid"

## Reliability and Validity



Reliable Not valid



Low validity Low reliability



Not reliable Not valid



Both reliable and valid