Utrecht University	CENTRE FOR DIGITAL HUMANITIES	19 September 2023
	Basics of Statistics	
	Session three  training for researchers and teachers in the Humanities	
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	Hypothesis testing	
	A hypothesis is a statement about a relationship between variables that can further be tested by a statistical model	
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## Hypothesis testing

Hypothesis testing in three steps

- Formulate hypotheses
   Compute a test statistic (some number that says something about group differences, for example)
   Compute the probability of that test statistic occurring under the null hypothesis (p-value)

Hypot	hesis	testing
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- Null hypothesis (H0)
  Typically, a "no effect" hypothesis
  There is no relationship between variables, no difference between groups, etc.
- Alternative hypothesis (HA)
  Statement that contains your prediction
  There is a relationship between variables, difference between groups, etc.

### Hypothesis testing

In statistical analyses, it is always  $\mbox{H0}\mbox{ that}$  is put to the test

The goal of hypothesis testing is rejecting H0 Of course, you never expect that there is nothing going on in the population, that is, that the null hypotheses is true.

So why do we test it?

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## Hypothesis testing

"The reason that we need the null hypothesis is because we cannot prove the alternative hypothesis using statistics, but we can reject the null hypothesis.

If our data give us confidence to reject the null hypothesis, then this provides <u>support</u> for our experimental hypothesis."

A. Field (2009) p.27

	Hypothesis testing		
	Two types of wrong decisions		
	Type I error You reject H0 when you should not		
	There is no effect in the population, but your sample data suggests so The probability of making a Type I error is denoted by alpha (o)	•	
	Type II error You do not reject H0 when you should There is an effect in the population, but your sample data do not reveal it The probability of making a Type II error is denoted by beta (§)		
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	Significance	-	
	<ul> <li>The p-value is the probability that you reject the null hypothesis while it is true (Type I error)</li> <li>The probability level at which we reject the null hypothesis is called the α-level</li> <li>The conventional value for α = 0.05 (Fisher, 1925)</li> <li>For an α-level of 0.05, you allow a 5% chance that you will falsely reject H0</li> <li>When we find a p-value &lt; 0.05 we reject the null hypothesis, and we say we have found a statistically significant effect</li> </ul>		
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	Significance		
	<ul> <li>In normal English, 'significant' means 'important'</li> <li>In statistics 'significant' means 'probably true' or 'not due to chance'</li> </ul>		
	A research finding may be true without being important / relevant		
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	Significance	-		
	Finding a significant effect is not sufficient	-		_
	<ul> <li>Significance (actually the p-value) is dependent on sample size</li> </ul>	-		
	<ul> <li>A large enough sample will almost always show a significant effect</li> <li>When you find a statistically significant effect, it does not</li> </ul>	-		
	necessarily mean that it's meaningful  How do you know your significant effect is also relevant?	-		_
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	Effect size	-		
	Liter 3/20	-		
	Effect size is a numerical way of expressing the strength or magnitude of an effect	-		
	<ul> <li>Amount of difference standardized to amount of dispersion</li> <li>The reporting of effect sizes facilitates the interpretation of the substantial, as opposed to the statistical, significance of a</li> </ul>	-		
	research result	-		
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	Effect size	_		
	Effect sizes are calculated on a common scale (standardized), which allows you to compare the effect of:	_		
	Different variables within a study     Same variables across studies     The reporting of effect sizes facilitates the interpretation of			
	the substantial, as opposed to the statistical, significance of a research result			
	Effect sizes complement statistical hypothesis testing     Effect sizes expresses the relevance of an effect	-		

Statistical power	
The probability that the statistical test correctly rejects the null hypothesis when the alternative hypothesis is true	

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# $\begin{tabular}{c|cccc} \textbf{Statistical power} \\ \hline \textbf{Truth about population} & H0 is true & Reject H0 & Fail to reject H0 & Type I error & Correct decision & Type II error & Type II error & Type II error & Type II error & Total Error & Tota$

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# Relationship between significance, effect size and power

- Significance and power are dependent on sample size
- Increasing sample size increases the chances of finding significant results, but it will not increase effect size
- However, increasing sample size will lead to more accurate estimates (=power), i.e. chances to find subtle effects (small effect sizes) will increase

# Relationship between significance, effect size and power

- Larger effects are easier to detect than smaller effects
  The smaller the effect you're looking for, in terms of significance, the more data you need to show it
  If your study lacks power, chances are high that you will not be able to detect an effect that is present in the population (Type II error)