

# 1 - Basic statistical concepts and data exploration

Describe the five steps of the research cycle.

## Empirical research cycle

1) Observation - Start with a question

An observation is not enough to conclude anything - Biases intrinsic to humans, no control group

The question defines the scope of the investigation

2) Generate theory - Makes predictions about future observations

Serves as a framework for your statistical model

3) Generate hypothesis

Specific and falsifiable

4) Collect data

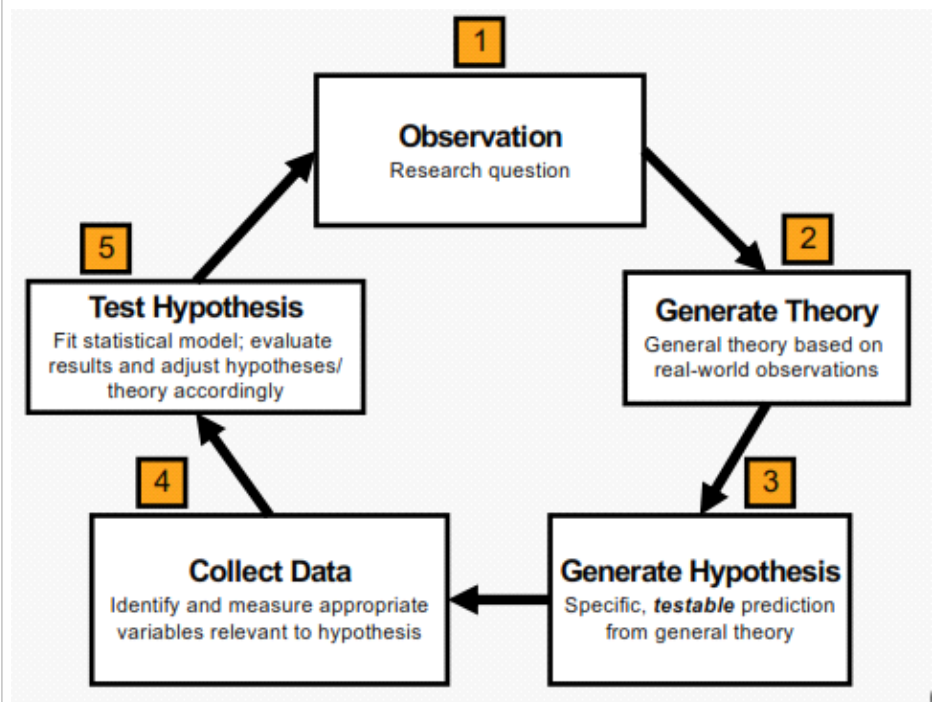
Increase or decrease confidence in the theory

No single data set can confirm or disprove a theory

4.A - Explore your data before analysing it (data entry is prone to error)

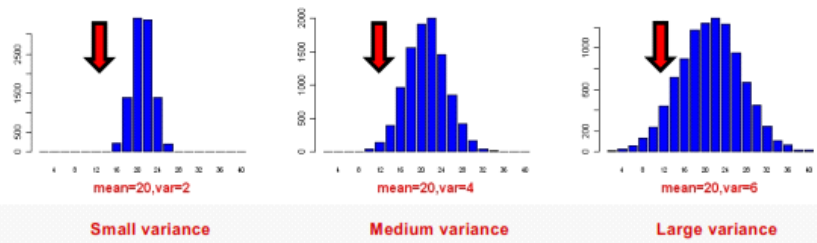
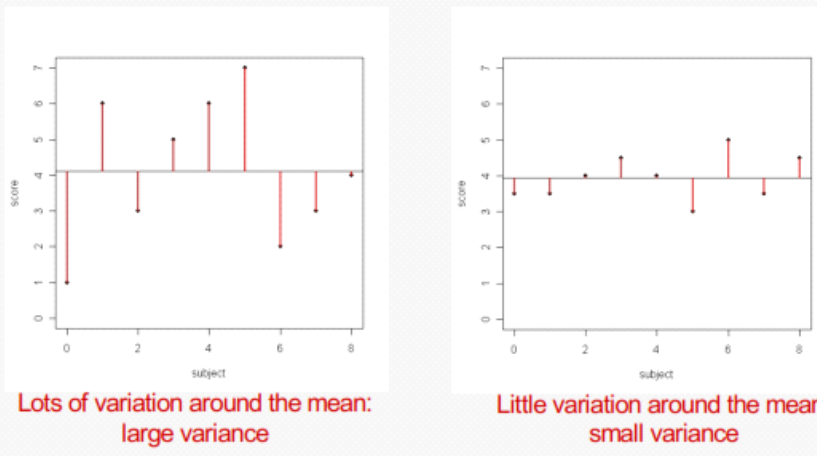
Graphs and frequency distributions (histograms)

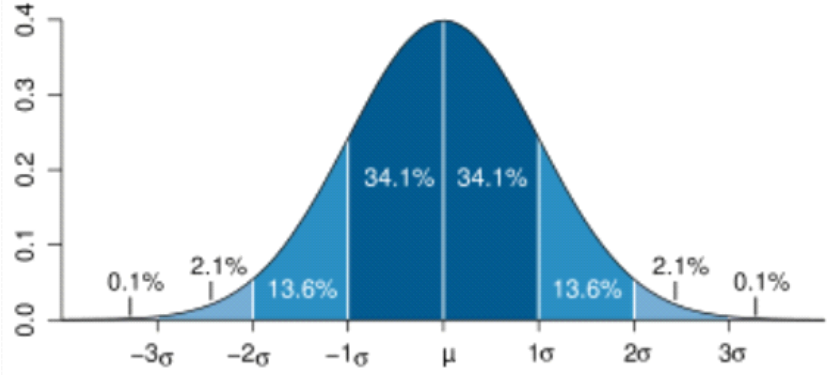
5) Test hypothesis - Is there a significant association between two or more variables

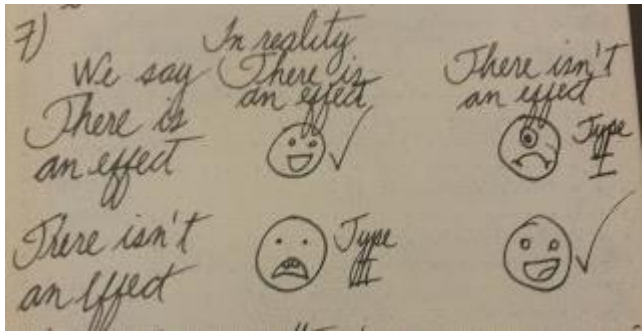


Example:

	<p><b>Research question:</b> Do boys and girls differ with respect to hair color?</p> <p><b>Theory:</b> If hair color genes are on the X chromosome, sex differences would be expected</p> <p><b>Hypothesis: (2-sided)</b> H<sub>0</sub>: Boys and girls <i>do not differ</i> in the proportions with each hair color H<sub>A</sub>: Boys and girls <i>differ</i> in the proportions with each hair color</p>
What are the two types of study design? What is their main difference?	<p><b>Types of study design</b></p> <p><b>Non-experimental:</b> Observational, non-invasive, cannot infer causation <b>Experimental:</b> Manipulation of variables, can infer causation</p>
What is a dataset?	<p><b>Variable:</b> Things that vary in a scientific investigation <b>Dataset:</b> Multiple variables</p>
What are the two variable types?	<p><b>Types of variables:</b> <b>Independent</b> - Predictor (cause) <b>Dependent</b> - Outcome (effect)</p>
What are the types of qualitative data?  What are the types of quantitative data?	<p><b>Measure data</b> <b>Qualitative:</b> Categorical (nominal or ordinal) <b>Quantitative:</b> Continuous (ratio or interval)</p> <p><b>Interval</b> - You can add and subtract, but not multiply. There is no absolute zero <b>Ratio</b> - There is an absolute zero</p> <p>All data could be measured in a discrete or continuous way Numbers in categorical data are arbitrary</p> <p>Measurement errors should always be considered in the data interpretation</p>
What does sum of squares, variance and standard deviation describe?  What is range?	<p>Descriptive statistics <b>Central tendency</b> Mode (nominal) - Highest frequency Median (ordinal) - Middle value after ranking Mean/average</p> <p><b>Dispersion</b> <b>Variance</b> - Mean deviation from the mean Needs to be square in order not to be zero <b>Standard deviation</b> - Square root of variance N - 1 due to degree of freedom (you already know the mean)</p>

	 <p><b>Range</b> - Distance between lowest and highest values  <b>Percentiles</b> - Each score relative to a standardized distribution (0-100)</p>
<p><b>What is the simplest statistical model?</b></p>	<p><b>Statistical models</b> - Describe your data well (minimize the error)  Outcome = model + error</p> <p>Simplest model = mean + standard deviation/error</p> 
<p><b>What is systematic and unsystematic variation?</b></p>	<p><b>Variance</b> - Describes the spread of distributions  <b>Systematic</b> - Due to known/manipulated factors  <b>Unsystematic</b> - Due to unknown/random factors</p>
<p><b>How do you know a sample is representative?</b></p>	<p><b>Sampling theory</b> - If we could collect all data, we would not need inferential statistics  Sample estimates represent population parameters - There is an inherent variability inherent in sampling</p> <p>How do you know a sample is representative?  Standard error = Variation in the sample means  When it is large enough (over 30), sampling distribution is normal</p> <p>SD - Within 1 samples  SE - Within multiple samples in the population</p>
<p><b>What is a test statistic?</b></p>	<p><b>Test statistic</b> - Ratio of systematic to unsystematic variance</p> <p>Proportions between groups - Chi-squared  Linear association</p>

<p><b>What are the parameters of a standard normal distribution</b></p>	<p>Comparison of means</p> <p>Distribution of the data</p> <p>Parameters of a standard normal distribution</p> <p>68% within 1 SD 95% within 2 SD 99.7% within 3 SD</p> <p>Mean = 0 SD = 1</p>  <p>Z value = 1.96 for 95% confidence interval</p>
<p><b>What are degrees of freedom?</b></p>	<p>Degrees of freedom - The amount of information you have to calculate a test statistic while accounting for the amount of information you are trying to estimate</p>
<p><b>What is the p-value?</b></p> <p><b>What is alpha?</b></p>	<p>Null hypothesis: H0 Alternative hypothesis (usually two-sided): H1</p> <p>What should the data look like if H0 was true?</p> <p>P-value = If H0 was true, what is the chance that this observation occurred? Alpha = Usually .05</p> <p>P &lt; alpha = Our result is unlikely under H0 (reject null hypothesis) P &gt; alpha = Our result is likely under H0 (not reject null hypothesis) P = alpha?</p>
<p><b>What is beta?</b></p> <p><b>Differentiate type I and type II errors</b></p>	<p>Alpha = Chance to accept a not real effect as present (detect non-existing effect) Beta = Chance to dismiss real defect as absent (not detect an existing effect)</p> <p>Type I error = alpha Type II error = beta</p>



Effect size - Pearson's r or Cohen's d

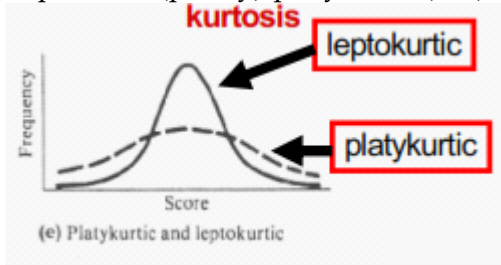
**Common assumptions**

- Parametric test
  - Observations are independent
  - Data is measured at a interval/ratio level
  - Normally distributed
  - Equal variance between groups
  
- Non-parametric test
  - Ordinal data

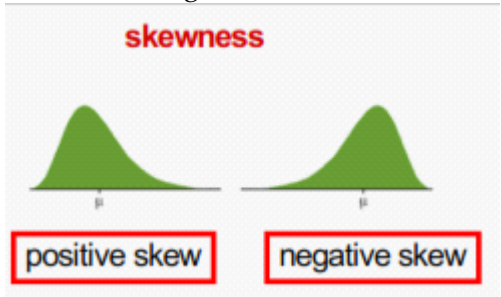
What is a leptokurtic curve? And a platykurtic?

**Skewness and kurtosis**

Leptokurtic(pointy)/platykurtic (flat)



Positive or negative skew



**What are outliers? What is their relation to the standard deviation?**

Outliers - Improbably large or small observations  
3 SD from the mean

**How to report study results**

- Design and sample characteristics
- N
- Mean
- SD

	Test statistic/p-value Effect size																
<b>What is the chi-squared formula?</b>  <b>How do you calculate the degrees of freedom in a chi-squared test?</b>	<p><b>Chi-squared test</b> - Difference in proportion in nominal variables          H0 - There is no difference between different groups</p> <p>Example: Colour hair in boys and girls</p> <p>Expected value = marginal values (total)</p> $\text{expected} = \frac{\text{marg}(a) * \text{marg}(b)}{\text{total}}$ <p>Example - Expected chance of a blonde boy (22 * 28 / 47)</p> <table border="1"> <thead> <tr> <th></th> <th>boy</th> <th>girl</th> <th>total</th> </tr> </thead> <tbody> <tr> <th>blond</th> <td>16 13.11</td> <td>12 14.89</td> <td>28</td> </tr> <tr> <th>red</th> <td>6 8.89</td> <td>13 10.11</td> <td>19</td> </tr> <tr> <th>total</th> <td>22</td> <td>25</td> <td>47</td> </tr> </tbody> </table> <p>Chi-squared formula</p> $\chi^2 = \sum \frac{(\text{obs} - \text{exp})^2}{\text{exp}}$ <p>Degrees of freedom = (number of rows - 1)*(number of columns - 1)</p> <p>Conditions for the test to be valid</p> <ul style="list-style-type: none"> <li>- No values = 0</li> <li>- Few values lower than 5</li> </ul>		boy	girl	total	blond	16 13.11	12 14.89	28	red	6 8.89	13 10.11	19	total	22	25	47
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Goals: Form opinions independent of others; formulate own research questions

Statistical techniques are stable over time

You have to think about statistics before designing the experiment

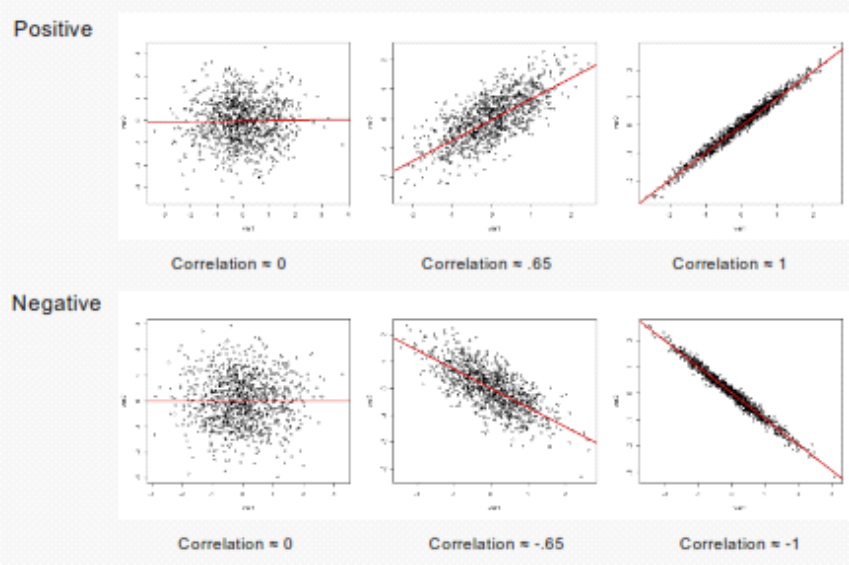
6 Sessions - Lectures, practicals, assignments, self-study  
 You have to pass all the assignments to be able to do the exam

Exam - Multiple choice and open-ended  
 Presentation - 10% of the grade

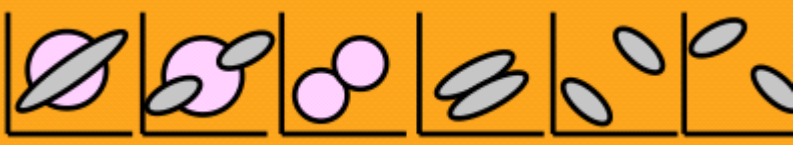
**Review week 1**  
 Sampling theory - Inferential statistics

- We cannot collect data about the entire population
- If we get enough samples - The sampling distribution will be a normal curve
  - Sampling mean = Standard error (standard distribution divided by the squared root of the sample)
- Sampling theory applies to other parameters (e.g. difference between populations, linear associations)
- Central limit theorem - If you have a large sample, it is more likely to be representative of the population (it does not mean that the population itself has a normal distribution)
  - Small samples are more likely not to be representative of the population (due to random chance)
- Test statistics - Effect size (ratio of systematic to unsystematic variability)
  - Sampling variability = effect size, express uncertainty (Z, t, F)
- Confidence intervals - Known distributional properties used to express uncertainty around an estimate
  - Z-score =  $x - \text{mean}$  divided by standard deviation
- Bootstrapping confidence intervals - Construct your own confidence intervals based on your own data
  - 1000 subsamples is the standard
  - Cut off top and bottom 2.5%
  - Remaining highest and lowest values are bootstrapped confidence
- Kolmogorov-S and Shapiro-W - Difference between your distribution and a theoretical distribution
  - KS - Any theoretical distribution
  - SW - Only normal distributions (more reliable for small and medium samples)
- Chi-squared - Margins take initial differences into account

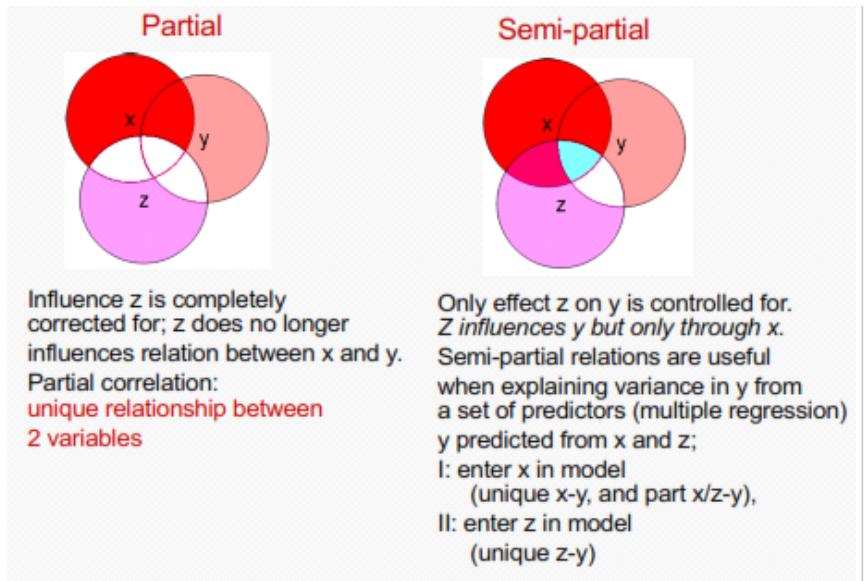
## 2 - Correlation and Regression

<p><b>What is the main difference between correlation and regression?</b></p>	<p>Linear association between two or more variables            Correlation - Non-directional            Regression - Direction</p> <p>They have more information than a pure means model</p>
<p><b>Why correlation does not imply causation?</b></p>	<p>Correlation - Strength and direction of variation between two variables            Values between +1 and -1</p>  <p>Does not equal causation - Third cause</p>
<p><b>How to calculate Pearson's correlation?</b>  <b>When to use, instead of Spearman?</b></p> <p><b>How can you make the correlation result not scale dependent?</b></p>	<p>Pearson's correlation rho            Standardized covariance</p> <p>Calculation:            Same as variance, but multiplying x by y instead of squaring it</p> $Covariance(x, y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{N - 1}$ <p>Scale dependent - Cannot compare studies unless they are using the same units of measure</p> <p>Solution - Divide this by standard deviation (correlation coefficient, not scale dependent)</p> $z = \frac{x - \bar{x}}{s}$ <p>Values are between +1 and -1</p>
<p><b>What is the difference between R and R squared?</b></p>	<p>Correlation significance            R is a standardized effect - Not a normal distribution (converted Z or t test)            R squared is an indication of how much variability in one</p>



	variable is explained by another																																																
<b>What are the assumptions for Pearson correlation?</b>	<p>Assumptions for Pearson</p> <ul style="list-style-type: none"> <li>Relations are linear</li> <li>Variables are bivariate normally</li> <li>Homoscedasticity of variances (similar variances between samples)</li> </ul>																																																
<b>How to calculate Spearman's correlation? When to use it over Pearson's?</b>	<p>Spearman's correlation rho</p> <ul style="list-style-type: none"> <li>Two ordinal variables or one ordinal and one interval/ratio</li> <li>Does not use scores, but ranks = Scale independent</li> </ul> <p>Calculation:</p> <ul style="list-style-type: none"> <li>Differences between ranks</li> <li>Squared differences between ranks</li> </ul> <div style="border: 1px solid gray; padding: 10px;"> <p><i>Rank scores within variables</i></p> <table border="1" style="margin-left: auto; margin-right: auto;"> <thead> <tr> <th>educ</th> <th>IQ</th> <th>Educ_r</th> <th>IQ_r</th> <th>diff</th> <th>diff<sup>2</sup></th> </tr> </thead> <tbody> <tr><td>5</td><td>121</td><td>5</td><td>6</td><td>1</td><td>1</td></tr> <tr><td>3</td><td>104</td><td>3</td><td>3</td><td>0</td><td>0</td></tr> <tr><td>2</td><td>78</td><td>2</td><td>1</td><td>1</td><td>1</td></tr> <tr><td>4</td><td>107</td><td>4</td><td>4</td><td>0</td><td>0</td></tr> <tr><td>1</td><td>85</td><td>1</td><td>2</td><td>1</td><td>1</td></tr> <tr><td>6</td><td>145</td><td>6</td><td>7</td><td>1</td><td>1</td></tr> <tr><td>7</td><td>120</td><td>7</td><td>5</td><td>2</td><td>4</td></tr> </tbody> </table> <math display="block">\rho = 1 - \frac{6 \sum \text{diff}^2}{n(n^2 - 1)} = 1 - \frac{6 * (1+1+0+0+1+1+4)}{7(49 - 1)} = 1 - .143 = .86</math> </div>	educ	IQ	Educ_r	IQ_r	diff	diff <sup>2</sup>	5	121	5	6	1	1	3	104	3	3	0	0	2	78	2	1	1	1	4	107	4	4	0	0	1	85	1	2	1	1	6	145	6	7	1	1	7	120	7	5	2	4
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7	120	7	5	2	4																																												
<b>What is a common correlation hazard?</b>	<p>Correlational hazards</p> <p>Combining subgroups in one sample is dangerous</p>  <p>Significant correlations do not imply causality and direction of causation (tertio quid)</p>																																																
<b>What is the difference between partial and</b>	<p>Partial correlations</p> <p>Correlation corrected for relationship with a third variable</p>																																																

semi-partial correlations?



Other correlations

- Kendall's tau - Small samples and many ties between differences
- Bi-serail - Dichotomus variable
- Polychronic - Two dichotomus variables

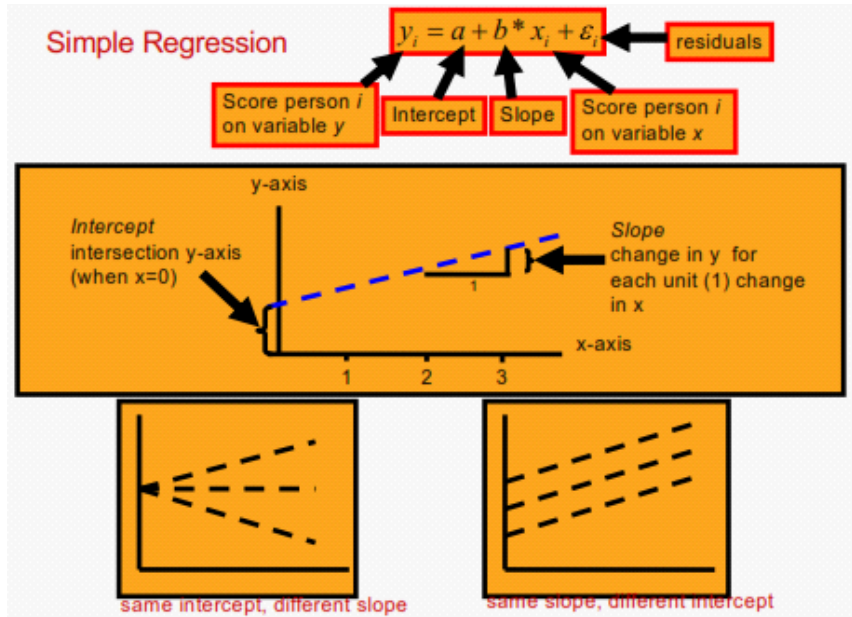
What can a regression do that a correlation cannot?

Regression

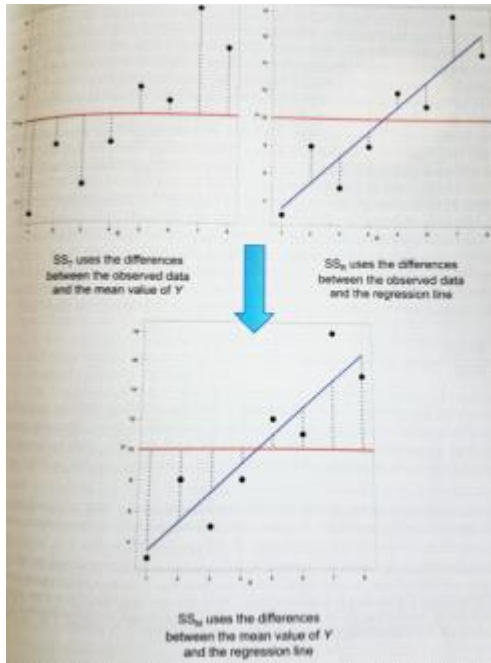
- Similar to correlation, but directional (one variable predicts another)
- It is a predictive model

What each variable means in the linear model?

Two bits of information are sufficient to describe a linear model  
 $Y = ax + b$



Best model minimizes total squared error (least squared estimate)



**Why should we use linear models?**

Why use a linear model?  
 Simple model - More parsimonious  
 You can predictive new values of the data

**What is the difference between simple and multiple regression?**

Simple regression - Two variables (one predictor and one outcome)  
 Multiple regression - More than two variables

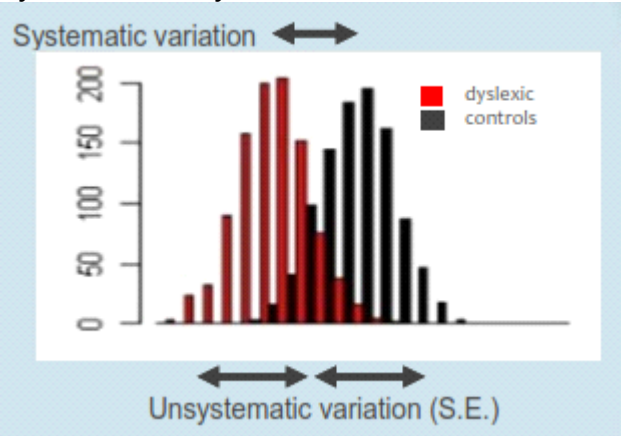
**Why transforming variables does not affect the interpretation of a regression?**

Transforming variables - Does not change the model or the interpretation of the data  
 Insensitive to linear transformations  
 Sensitive to non-linear transformations (e.g log)

**Under which circumstances are correlations and regressions the same?**

Standardizing both variables = Correlation and regression is the same!  
 $B = \text{correlation } r$   
 $A = 0$

### 3 - Between subject designs

<p><b>What is the difference between within subject design and between subject design?</b></p>	<p><b>Within subject design</b> - Dependent, overlapping samples (same subjects are tested more than once)  <b>Between subject design</b> - Independent, non-overlapping samples</p>
<p><b>When should you use a t-test?</b></p>	<p><b>Independent sample t-test</b>          Two groups - Control and experimental groups          Assumption: Difference in means can be attributed to the manipulation (causal interpretation)          Random assignment</p>
<p><b>What is the difference between experimental and quasi-experimental design?</b></p>	<p><b>Experimental vs quasi-experimental</b>  <b>Experiment</b> - Experimental factor is manipulated by reasearcher, random assignment, control group, equivalent control and experimental group (match)  <b>Quasi-experiment</b> - Experimental factor not manipulated by researcher, no random assignment, no real control, no match between experimental and control group</p>
<p><b>What is the question that a t-test tries to answer?</b></p> <p><b>How can systematic and unsystematic variation be observed in a distribution?</b></p>	<p>Question of t-test: <b>Do this data come from the same population or from difference population?</b>          Answer depends on the variance</p> <p><b>Systemic vs unsystematic variation</b></p>  <p><b>Systematic</b> - Due to the fact that the groups are different  <b>Unsystematic</b> - The variance between the same group (due to other factors outside from the experimental manipulation)</p>
<p><b>What the formula for a t-</b></p>	<p>Formula for t-test</p>

test?

Observed difference between means (systematic variation)

Expected difference under H0:  $\mu_1 - \mu_2 = 0$

$$t = \frac{(X_1 - X_2) - (\mu_1 - \mu_2)}{SE} = \frac{(X_1 - X_2)}{SE} = \frac{D}{SE}$$

Unsystematic variation: Variation observed by chance alone

(difference between observed means) - (difference between expected means)/standard error

Systematic variation/unsystematic variation - Like all test statistics

Under H0 - the difference between **expected means is 0**

Calculation of standard error

If samples are equal

$$SE = \sqrt{\frac{s_1^2}{N_1} + \frac{s_2^2}{N_2}}$$

If samples are not equal

$$SE = \sqrt{\frac{s_p^2}{N_1} + \frac{s_p^2}{N_2}}, \text{ where}$$

$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{(n_1 + n_2 - 2)}$$

**Why should you report an effect size along with a t-test?**

**Effect size** - Quantitative measure of the magnitude of a phenomenon, allows comparisons between different studies

$$Df = N - 2$$

$$r = \sqrt{\frac{t^2}{t^2 + df}}$$

R = .1 (small effect)

R = .3 (medium effect)

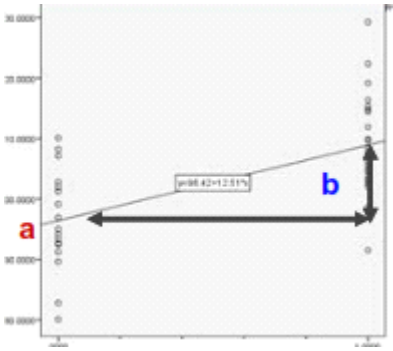
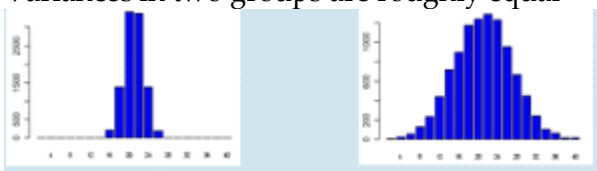
R = .5 (large effect)

It is always good to calculate the effect size and report it along with the t-test

**How can you maximize the probability of observing a significant experimental effect?**

How can you maximize the probability of observing a significant experimental effect?

- Increase sample size
- Decrease variance
- Increase effect size/mean difference (the difference between the means between two groups)

	<p>It is important to think about this before collecting data (during the experimental design)</p>
<p><b>What is the interpretation of a and b in a regression line?</b></p>	<p>Analogy t-test - regression  A = Mean value of Y when group variable is 0  B = Mean value of Y with 1 increase in group variable  B is the difference between the two means</p>  <p>Examples  Overall mean  Means between groups</p> <p>Interpretation of a - The difference between the first group and 0 is significant  Interpretation of b - The difference between groups 1 and 2 is significant</p>
<p><b>What are the assumptions of a t-test</b></p>	<p><b>Assumptions of t-test</b>  Data are normally distributed  Data are measured at interval/ratio level  Variances in two groups are roughly equal</p>  <p>Probably there was some problem in the experimental design if homoscedasticity is not met</p> <p>Observations are independent - Come from different subjects, not correlated between groups</p>
<p><b>What a significant Levene test indicates?</b></p>	<p>Levene test less than .05 - Variances are not equivalent between groups  Then you can use equal variances not assumed</p>
	<p>How to report a t-test</p>

The arithmetic capacity of children who experience reading disability (RD, N=24, M=.03, SD=.79) was compared to that of chronological age controls (CA, N=24, M=.07, SD=.37). Because the variances of the two groups were significantly different ( $F(1,46)=10.05, p < .01$ ), a t-test for unequal variances was conducted. This test was not significant ( $t(32.71)=-.54, ns, r = .09, \text{Cohen's } d=.26$ ). In other words, no difference in arithmetic capacity between RDs and CAs could be detected.

When should you use an one-way ANOVA?

**One-way ANOVA**

Compares **more than two means**

When you manipulate one or multiple variables - e.g. Control, KO, double KO

One dependent variable = one independent variable with 3 or more levels

Why dummy coding is used in an one-way ANOVA?

Explained in terms of regression

$Y = a + bG + b_2G_2$  (dummy coding)

**Three groups: multiple regression**

$$y_i = a + b_1G_{i1} + b_2G_{i2} + \epsilon_i$$

Group 1:  $y_i = a + \epsilon_i$   
 Group 2:  $y_i = a + b_1 + \epsilon_i$   
 Group 3:  $y_i = a + b_2 + \epsilon_i$

"Dummy coding"

	Group1	Group2	Group3
G1	0	1	0
G2	0	0	1

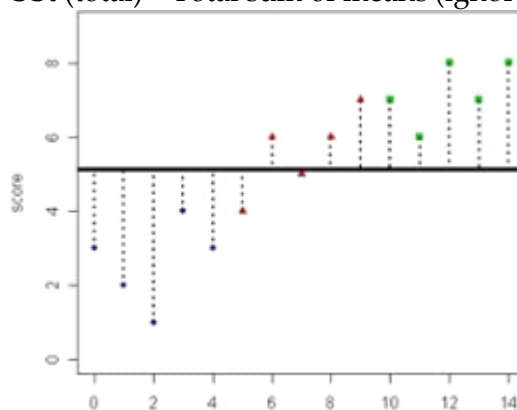
Each level of the dichotomous variable is contrasted to a reference level

Define SSt, SSm and SSr.

Means for each group

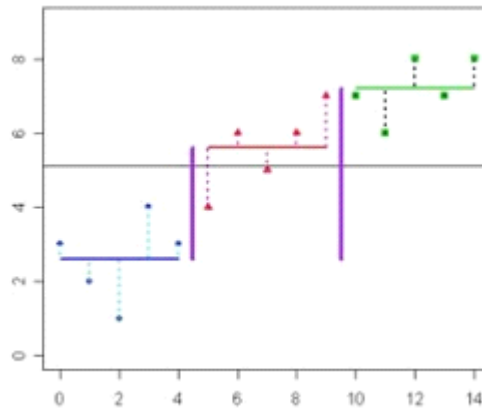
**SSt** (total) = Total sum of means (ignores group)

What is the formula of a f-statistic?



**SSm** (model) = Total sum of mean (taking groups into account)

If this is different than SSt, it means that group membership is sufficient to explain variation in the data



$SSr$  (residual) = Residual variation unexplained by  $SS_t$  and  $SS_m$

F-statistic =  $SS_m/SSr$  (what is explained by the model divided by what is not explained by the model)

**Define the three different types of planned contrast in an one-way ANOVA.**

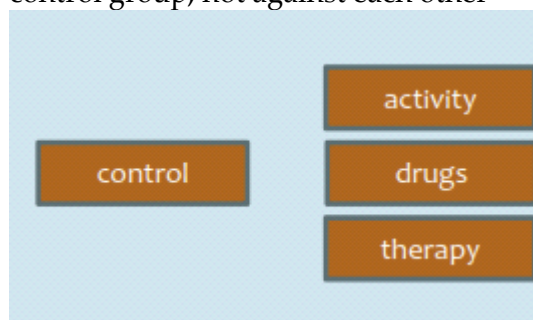
F-ratio more than 1 - More systematic than unsystematic variation

Under  $H_0$  - Systematic variation should be zero

If F-ratio is significant, we know there is a difference, but we do not know how (do all groups differ from each other? Only one has a significant difference?)

Planned contrast - Often one-sided

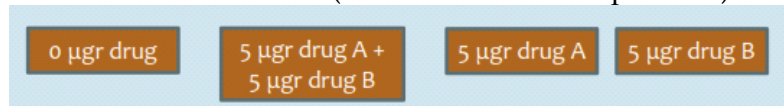
Simple - You compare experimental groups against the control group, not against each other



Repeated - Sequentially compare groups in a logical order (severity manipulation)



Helmert - A+B vs A or B (combination of manipulation)



In SPSS - Analyse->GLM

Post-hoc - Often two sided

**Why is ANOVA always a two-sided test?**

ANOVA: 1 or 2 sided?

Either all means are equal or they are not equal - ANOVA is always 2-sided



	<p>Post-hoc tests - Can be one-sided</p> <p>Manipulation groups in depression - A manipulation cannot increase depression levels</p>
<p><b>What are the assumption for an ANOVA</b></p>	<p>Assumptions for ANOVA</p> <p>Dependent variable is normally distributed</p> <p>Equal variance in each group</p> <p>Independence observations</p>
<p><b>When should you use a factorial ANOVA?</b></p>	<p><b>Factorial ANOVA (2-way, 3-way)</b></p> <p>Multiple means and multiple predictors</p> <p>E.g. Control vs drug; Home vs clinique; male vs female</p> <p>Two-way ANOVA - Diet1/Diet2/no diet; male/female (Image)</p> <p>Interpretation</p> <p>Main effect diet - Across two gender groups</p> <p>Main effect sex- Across all dietary conditions</p> <p>Interaction diet vs sex - The effect of manipulation is the same in all groups?</p>
<p><b>How can you visually determine if there is an effect of the manipulation, an effect of the group and an effect of the interaction?</b></p>	<p>Interpretation interaction</p>
<p><b>When should you use a non-parametric equivalent of ANOVA?</b></p>	<p>Parametric vs non-parametric</p> <p>Parametric - Assumptions about distribution of the data, for interval and ratio data</p> <p>Non-parametric - For data with outliers that you do not want to</p>

remove; N is small - If your distribution is normal, some argue that you should use a t-test

## 4 - Within subject design

**When should you use paired sample t-test over normal t-tests?**

Within subject - Multiple measures on the same subject

Independent sample t-test

Systematic variation - Caused by the manipulation

Unsystematic variation - Not caused by the manipulation  
(observations at different times, genetic variability of mice)

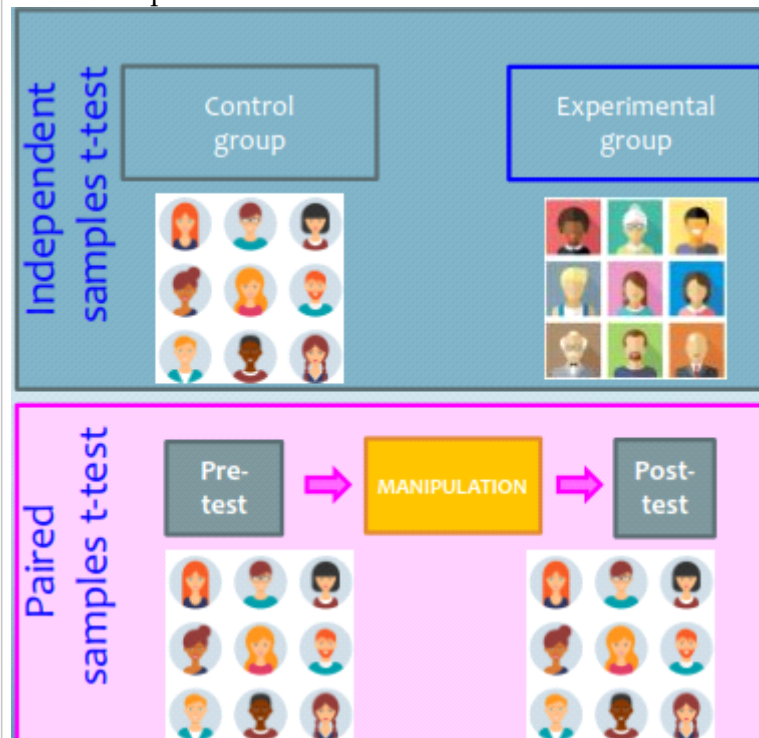
Cohen's  $d$  = Effect size for mean difference between two groups

T-test cannot be used when the samples are not independent

**Under which circumstances should paired samples t-test be used?**

Paired samples t-test

Same people in every condition -> Change is due to a manipulation



*Difference between independent and paired samples t-tests*

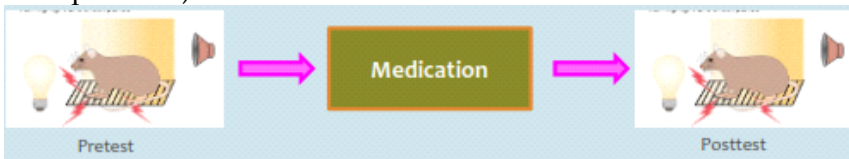
### Exemples of experimental designs

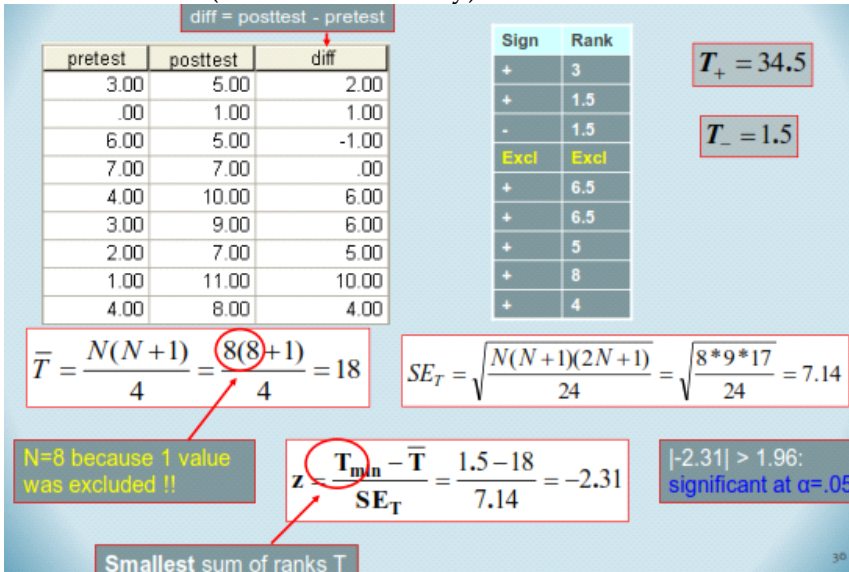
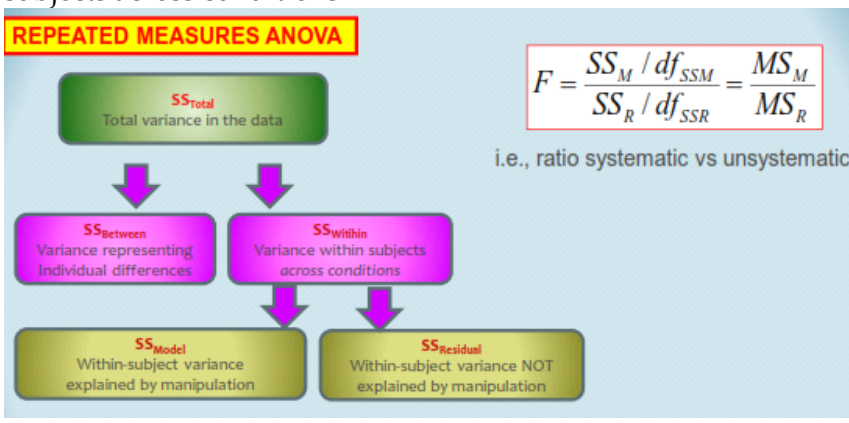
**Pairs of cases** - Placebo and drug group with match background (age, gender, height, blood pressure, insuline levels)

Matched pair parallel design

Pairs of cells from the same brain region

One has a manipulation, one has not

	<p><b>Pre-test -&gt; X -&gt; Post-test</b>          Training, Intervention, Medication, Passing of time</p> <p>Paired t-test can be used for <b>different evaluations by different people</b>          Teacher and mother evaluation of student musical performance</p>
	<p>Quasi-experimental designs          WT and KO pups from the same litter (womb environment is the same)          Pairs of males and females matched on IQ, educational background, ethnici</p>
	<p>Calculation</p> <div style="border: 1px solid #ccc; padding: 10px; background-color: #e0f0ff;"> <p>Mean of observed differences between conditions 1 and 2</p> <p>Expected difference under H0</p> <math display="block">t = \frac{\sum(x_{i1} - x_{i2})/N - D_{exp}}{SE_D} = \frac{D_{obs} - D_{exp}}{s_D/\sqrt{N}} = \frac{D_{obs}}{s_D/\sqrt{N}}</math> <p>SE of the differences</p> <p>DF=N-1</p> <p>Main difference with independent samples t-test: calculation of mean(D)</p> </div> <p>Mean of observed differences between conditions 1 and 2 divided by the standard error (df = n - 1)</p> <p>Cohen's d for paired samples</p> $d = \frac{\bar{X}_1 - \bar{X}_2}{SD} = \frac{t}{\sqrt{N}}$
<p><b>What are the main advantages and disadvantages of paired sample t-tests?</b></p>	<p><b>Advantages of paired t-test</b>          Perfect matching - More power to detect effect manipulation          Has a smaller unsystematic variation - Variation within the same person changes less          It is often one-sided, since you have a clear idea of the type of effect that the manipulation will have (divide alpha levels by 2)</p> <p><b>Disadvantages</b>          Depending on design - Carry-over effect (people learn from the manipulation)</p> <div style="text-align: center;">  <p>Pretest → Medication → Posttest</p> </div> <p>- The control group has to be exposed to the pre-test and post-test but without the manipulation!</p>
<p><b>What are the assumptions of paired</b></p>	<p><b>Assumptions of paired t-test</b>          Data is interval/ratio</p>

sample t-tests?	Differences between scores are normally distributed Observations have no carry-over effect																																																		
When should you use Wilcoxon's signed rank test instead of paired t-tests?	Non-parametric variant - <b>Wilcoxon's signed rank test</b> Data is ordinal Group is small Differences are not normally distributed																																																		
How is the 'difference' defined in Wilcoxon's signed rank test?	<p>Calculation of Wilcoxon's Based on ranks (like Mann-whitney)</p>  <p>diff = posttest - pretest</p> <table border="1" data-bbox="673 478 1031 755"> <thead> <tr> <th>pretest</th> <th>posttest</th> <th>diff</th> </tr> </thead> <tbody> <tr><td>3.00</td><td>5.00</td><td>2.00</td></tr> <tr><td>.00</td><td>1.00</td><td>1.00</td></tr> <tr><td>6.00</td><td>5.00</td><td>-1.00</td></tr> <tr><td>7.00</td><td>7.00</td><td>.00</td></tr> <tr><td>4.00</td><td>10.00</td><td>6.00</td></tr> <tr><td>3.00</td><td>9.00</td><td>6.00</td></tr> <tr><td>2.00</td><td>7.00</td><td>5.00</td></tr> <tr><td>1.00</td><td>11.00</td><td>10.00</td></tr> <tr><td>4.00</td><td>8.00</td><td>4.00</td></tr> </tbody> </table> <table border="1" data-bbox="1128 457 1274 755"> <thead> <tr> <th>Sign</th> <th>Rank</th> </tr> </thead> <tbody> <tr><td>+</td><td>3</td></tr> <tr><td>+</td><td>1.5</td></tr> <tr><td>-</td><td>1.5</td></tr> <tr><td>Excl</td><td>Excl</td></tr> <tr><td>+</td><td>6.5</td></tr> <tr><td>+</td><td>6.5</td></tr> <tr><td>+</td><td>5</td></tr> <tr><td>+</td><td>8</td></tr> <tr><td>+</td><td>4</td></tr> </tbody> </table> <p><math>T_+ = 34.5</math> <math>T_- = 1.5</math></p> <p><math>\bar{T} = \frac{N(N+1)}{4} = \frac{8(8+1)}{4} = 18</math></p> <p><math>SE_T = \sqrt{\frac{N(N+1)(2N+1)}{24}} = \sqrt{\frac{8*9*17}{24}} = 7.14</math></p> <p><math>z = \frac{T_{min} - \bar{T}}{SE_T} = \frac{1.5 - 18}{7.14} = -2.31</math></p> <p>N=8 because 1 value was excluded !!</p> <p> -2.31  &gt; 1.96: significant at <math>\alpha = .05</math></p> <p>Smallest sum of ranks T</p>	pretest	posttest	diff	3.00	5.00	2.00	.00	1.00	1.00	6.00	5.00	-1.00	7.00	7.00	.00	4.00	10.00	6.00	3.00	9.00	6.00	2.00	7.00	5.00	1.00	11.00	10.00	4.00	8.00	4.00	Sign	Rank	+	3	+	1.5	-	1.5	Excl	Excl	+	6.5	+	6.5	+	5	+	8	+	4
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How to report Wilcoxon's results?	How to report You report the median, not the mean! You are comparing ranks, not values Always report a z-score																																																		
When should you use repeated measures ANOVA?	Repeated measures ANOVA - <b>Analysis of within subjects across conditions</b> More than 2 measures in each subject - Plant growth over multiple days - Global temperature over time																																																		
	<p><b>Calculation</b></p> <p>Variation within subjects across conditions - Variance within subjects across conditions</p>  <p><b>REPEATED MEASURES ANOVA</b></p> <p><math>F = \frac{SS_M / df_{SSM}}{SS_R / df_{SSR}} = \frac{MS_M}{MS_R}</math></p> <p>i.e., ratio systematic vs unsystematic</p> <p><b>SS<sub>Total</sub></b> Total variance in the data</p> <p><b>SS<sub>between</sub></b> Variance representing individual differences</p> <p><b>SS<sub>within</sub></b> Variance within subjects across conditions</p> <p><b>SS<sub>Model</sub></b> Within-subject variance explained by manipulation</p> <p><b>SS<sub>residual</sub></b> Within-subject variance NOT explained by manipulation</p>																																																		

$$SS_W = \sum (x_{ik} - \bar{x}_i)^2 = s_i^2 (k - 1)$$

Score person i in condition k      Mean score person i      Number of conditions

Variance in person i


Define three types of Post hoc analysis to repeated measures ANOVA

Post hoc: contrasts  
 Simple - 1 vs 2, 1 vs 3, 1 vs 4  
 Repeated - 1 vs 2, 2 vs 3, 3 vs 4  
 Deviation - Mean level to overall grand mean  
 Helmert - Each level to later levels

Define sphericity.  
 Why the sphericity of an experiment with two groups cannot be calculated?

Sphericity - Differences between conditions  
 Determine by Mauchly's test (chi-square)  
 Homogeneity of the variance of the differences (variance1 = variance 2 = variance3)

**Sphericity**  
 In regular ANOVA: homogeneity of variance between groups



In repeated measures ANOVA we work with differences between conditions  
 → homogeneity of the variance of the differences

**3 conditions:**  
 variance<sub>1,2</sub> ≈ variance<sub>1,3</sub> ≈ variance<sub>2,3</sub>

*Sphericity cannot be done when there are only two groups!*

SPSS gives you r squared (overall effect size) = square root of partial Eta squared

Sig.	Partial Eta Squared
.000	.804
.000	.804
.000	.804
.000	.804

**Overall effect size**  
 $\eta^2 = r^2$ , so  
 $r \approx .90$

What is the main difference between factorial repeated measures and repeated measures ANOVA?

Factorial repeated measures  
 Within subject factor - Variable is manipulated within a subject - Time, dose, sessions  
 Between subject factor - Variable distinguishes different groups

Example - Diet  
 Main diet

Main gender

Interaction

A new diet is developed that should help people loose weight.

Randomly selected men and women are put on the diet and after a baseline measure, are measured another 4 times to see how their weight changes over time.

**Research questions**

**Main diet:** does weight change over time due to the diet?

**Main gender:** do men and women differ in their mean weight?

**Interaction:** is the effect of the diet the same in men and women?

What is missing from this design to make it a true experiment??

**CONTROL GROUPS!!**



t1 t2 t3 t4 t5

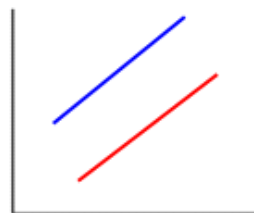
Each point has a research question (H0 and H1)

You always need a control group

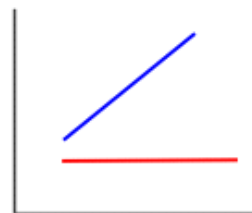
How can each of this effect be visually observed in a graph:

- a) Manipulation
- b) Group
- c) Interaction

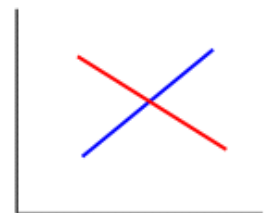
Understand graph interactions for exam



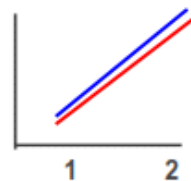
1 2  
Main manipulation  
Main group  
No interaction



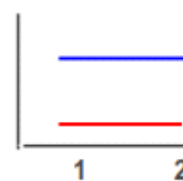
1 2  
Main manipulation  
Main group  
Interaction



1 2  
No main manipulation  
No main group  
Interaction



Main manipulation  
No main group  
No interaction



No main manipulation  
Main group  
No interaction

Manipulation - The mean of pretest and posttest is different

Group - There is a space between the lines

Interaction - The lines are going in different ways

When should Friedman's ANOVA be used instead of factorial repeated measures?

Friedman's ANOVA - Non-parametric version

- Ordinal data
- Small sample size
- Sphericity assumption violated
- Differences are not normally distributed

Does not allow the study of interactions

Only allows non-parametric post-hocs

When should the Mc Nemar test be used?

Mc Nemar test - Repeated measures for dichotomous data

E.g. Drinkers and non-drinkers

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Assignment review

Multiple groups - N, mean and standard deviation for each

Always check assumption before running the analysis

Report df, test statistics and p-value

F-test has two p-values - Report both (degrees of freedom of model and the distribution)

$SS_t = n - 1$

$SS_r = N - k$

$SS_m = k - 1$

Always explicitly state research question, H0 and H1



# 5 - Multiple regression

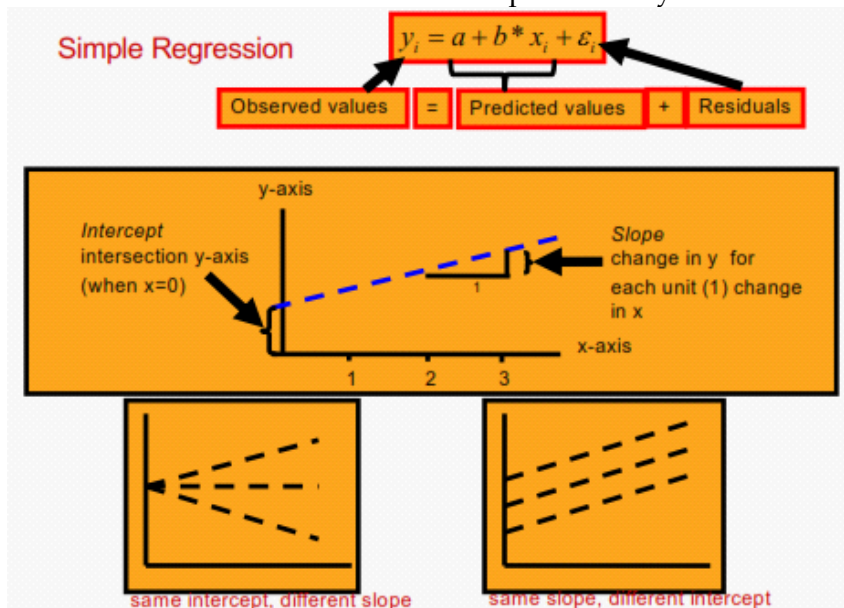
**What is a and b in a simple regression model?**

Simple regression - Describes a line

$$Y = a + bx$$

Explained variance in regression formulas

How much of the observed values are predicted by the model?



SSr - Variance of the regression model to the means model

**What is the criteria to assess if you should use a regression or not?**

Criteria for regression model evaluation

Hypothesis testing

Regression coefficients

Full model

Parsimony - Simple models with few predictors are preferred to complex models

**What does B represent in an one-side regression model? And in a two-sided regression?**

Two-sided test

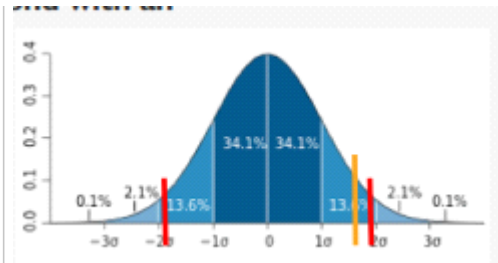
Null hypothesis - Slope (B) = 0

Alternative hypothesis - B different from 0

One-sided test

Null hypothesis - B <= 0

Alternative hypothesis - B > 0



**What does an F-test represent?**

F-test  
Ratio improvement of regression model compared to means model

$$F = \frac{MS_M}{MS_R} = \frac{SS_M / df_M}{SS_R / df_R} = \frac{(SS_T - SS_R) / df_M}{SS_R / df_R}$$

Explained vs non-explained

**Why is an adjusted R squared important to be reported?**

Example on SPSS  
Arithmetic ability based on Raven task?  
- Adjusted R squared - As you add more predictors to the model, r squared tend to increase on its own  
B - constant (a), intercept  
Beta coefficient - for every increase of x, how much is the increase of y (standardized in standard deviations)

How to report a regression

Arithmetic ability and IQ were measured in 170 children. Arithmetic ability was measured on a scale... ranging from... to ... (M<sub>a</sub>, SD<sub>a</sub>); Raven was measured on... (M<sub>iq</sub>, SD<sub>iq</sub>). Aim of the study: to investigate whether IQ significantly predicts arithmetic ability. To this end, a regression analysis was run with arithmetic ability as dependent variable, and IQ as predictor. The F-statistic was significant (F(1, 168) = 15.87, p < .001), implying that... Both the intercept and the regression weight for IQ were significant (t(168) = 6.03, p < .001, t(168) = 3.98, p < .001, respectively). The resulting equation thus looks as follows:

$$\text{Arithi} = 74.17 + 1.145 * \text{Raveni} + \epsilon_i$$

Correlation Raven-Arith was .29, and R<sup>2</sup> equalled .09 (adjusted R<sup>2</sup> was...), meaning that...

Inspection of the standardized residuals shows that less than 2% > |2.58|, and < 5% > |1.96|, implying that...

**What are the implicit and explicit assumptions to perform a regression?**

Implicit assumptions for regression

- Predictor is quantitative or categorical with 2 levels
- Dependent variable is continuous
- No perfect multicollinearity between predictors
- Predictors are not correlated with external variable

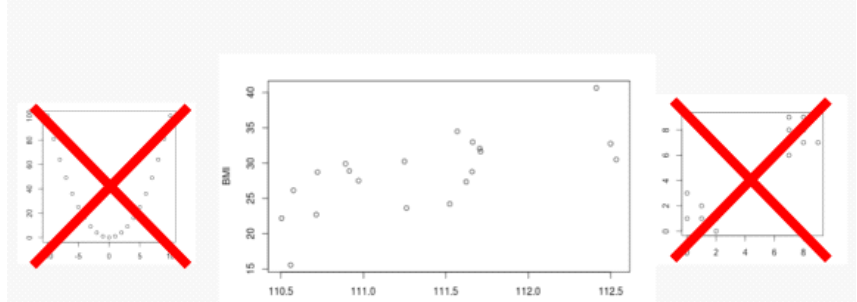
Explicit assumptions for regression

- Linearity - Always assess this with a scatter plot
- Independent observations - No repeated observations from the

same subject, no hierarchical organization in subjects/animals  
 Durbin-Watson test  
 Graph residuals can identify gross deviation  
 Homoscedasticity - Variation outcome should be constant across predictor range  
 Boxplot  
 Normal residuals - Mean equals 0  
 If violated: Non-parametric testing

Outliers can influence regression model

**Very simple to assess with a scatterplot – ALWAYS look at this before analysis**  
 If violated: transform variable(s) or use alternative (non-linear) regression model



- Checking for outliers is very important in regression analysis - Always make a scatterplot of the data

**If regression and correlation are the same thing, why do regressions at all?**

Why bother with regression?  
 Mathematically, regression and correlation are the same  
 Conceptually, you frame them differently (regression provides a scalable framework)

**What is a multiple regression?**

Multiple regression  
 More complex questions require more complex models  
 $Y = b_0 + b_1X_1 + b_2X_2 + b_3X_3$

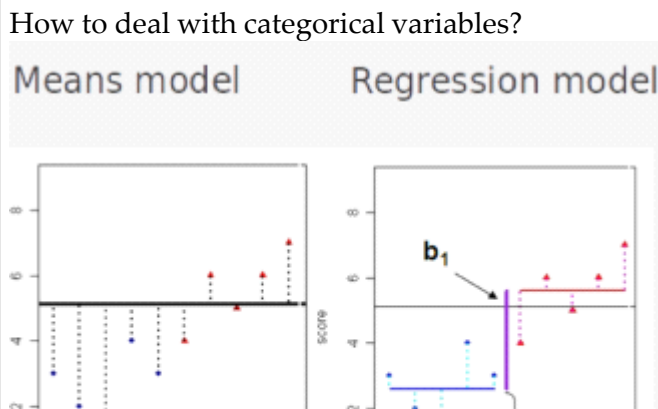
**What is R-squared?**

R squared - percentage of variability explained by all predictors

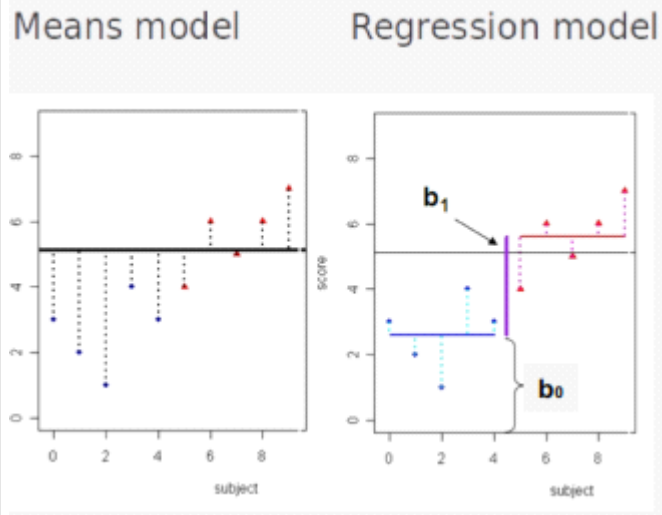
**What are the different types of predictors entry for multiple regressions?**

Enter mode - All predictors entered in the equation at once (no order)  
 Hierarchical regression - Order of predictors is chosen by experimenter  
 Stepwise - Decided by the computer (which generates largest R squared)  
 Forward - Improve R square by adding predictor  
 Reverse - Improve R square by removing predictors

**How to deal with categorical variables in a multiple regression?**



categorical variables in a multiple regression?



Dichotomous - Indicator variable (0 and 1)  
 T-test can be converted into a regression model  
 More than two variables- Dummy coding  
 Number of dummy categories equals number of levels minus 1

What is dummy coding?

What if effect of predictor depends on group  
 - Interaction is determined by dummy coding

Condition	X1	X2	X3
No intervention	0	0	0
Placebo	1	0	0
Old drug	0	1	0
New drug	0	0	1

Multiple continuous variables  
 Displacement measured as continuous variables

Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	33.403	2.837		11.776	.000
	number of cylinders	-2.340	.395	-.690	-5.919	.000
	transmission	3.331	1.463	.265	2.276	.031

a. Dependent Variable: miles per gallon

- Intercept ( $H_0: b_0 = 0$ )
- Main effects ( $H_0: b_1=0 / b_2=0$ )
  - Beta & Std. beta control for effects of other predictors, so coefficients will change when they are added/removed from the model
  - Remove non-significant predictors and re-evaluate final model

What are the assumptions of multiple

Assumptions for multiple regressions  
 Linearity

regressions?	<p>Independent observations  Homocedasticity  Normal residuals  No perfect multicollinearity</p>
<p><b>What is Variance Inflation Factor? From which value should you consider removing a variable?</b></p>	<p>Very large correlation between covariates is problematic</p> <p><b>Variance Inflation Factor (VIF)</b> is measure to estimate how redundant each covariate in your regression is</p> <ul style="list-style-type: none"> <li>It performs a regression with one covariate as outcome and all other covariates as predictors [<math>VIF = 1 / (1-R^2)</math>]</li> <li>VIF = 1: uncorrelated with all other covariates in the original regression (<math>R^2=0</math>)</li> <li>VIF &gt; 5: very correlated; consider removing this variable (<math>R^2&gt;0.8</math>)</li> <li>VIF &gt; 10: extremely correlated; you probably should remove this variable (<math>R^2&gt;0.9</math>)</li> </ul> <p>Each predictor has its own VIF; if necessary remove variable with highest VIF and recompute.</p>
<p><b>What is a familywise error?</b></p>	<p>The more tests you do, the more likely you are to get a false positive  Setting alpha (probability to reject the null hypothesis even though it was true = type I error)  Familywise error = <math>1 - 0.95(n)</math>  10 tests = 40%</p> <p>In neuroscience, many designs are characterized by many tests  Proteomics  Genomics  Metabolomics</p>
<p><b>What are the classes of correction for familywise error?</b></p>	<p>Classes of correction  Family-wise error rate (FWER)  False discovery rate (FDR)  False discovery proportion (FDP)</p>
<p><b>What is the difference between Bonferroni correction and Holm correction?</b></p>	<p>Familywise error rate -  Bonferroni correction - Divide alpha by the number of tests  Often too conservative when p-values are correlated</p> <p>Holm - Sort the p-values sequentially, than define the critical value for each p-value (<math>\alpha_i = \alpha / (m-i+1)</math>)  Less conservative in Bonferroni</p> <p>Tukey - Similar to Holm (put p-values in order)</p>
<p><b>What is false discovery rate?</b></p>	<p>False discovery rate  FDR - Hits by chance alone and hits observed -&gt; Ratio</p> <p>Benjamini &amp; Hochberg  Read Verhoeven (2005)  'm' - Control for the number of hypothesis</p>

Errata - Effect size (Z-score/square root of N)

N is the total number of observations in the study

Normality tests - Always report df, test statistic and p-value

Always report the specific p-value

4.1.

The two hypothesis (null and alternative) need to cover all the possible outcomes

The **differences** should be normally distributed

Rsquared =

Always report effect size

4.2

Weight of three strains of mice

- Main weight - The weight changes across time
- Strain - Do the strain differs in weight
- Interaction - Do the strains gain weight in different ammounts

Always report all the effects

You don't do a post-hoc test if your effect is not significant

4.3

DBP in increasingly stressful conditions

Always report the groups, time points, what is being tested (N, median, sd)

Test the normality of differences

Friedman ANOVA is a chi-square and needs to be reported as such

Always report effect size

Bonferroni - Related samples

Tukey - Unrelated samples

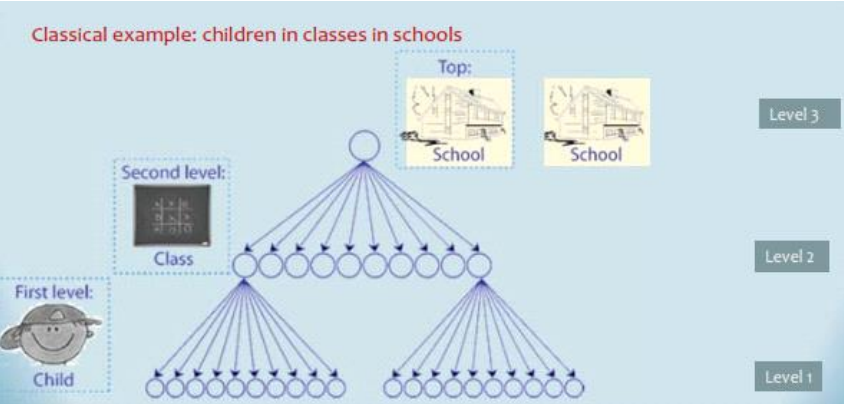
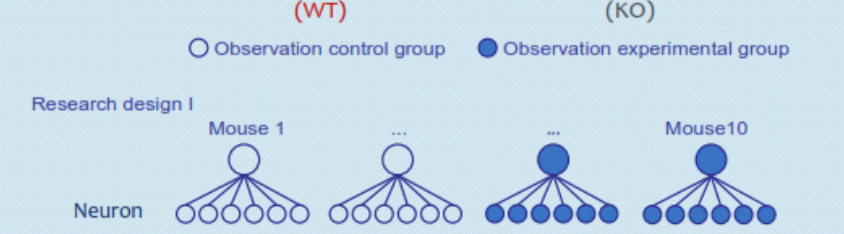
1 or 2-sided question

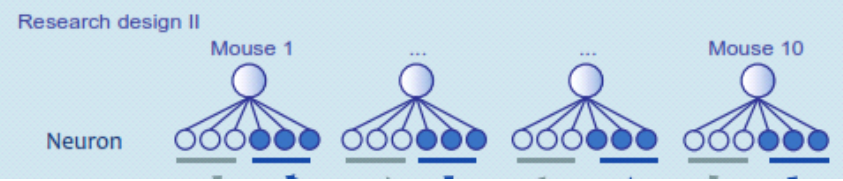
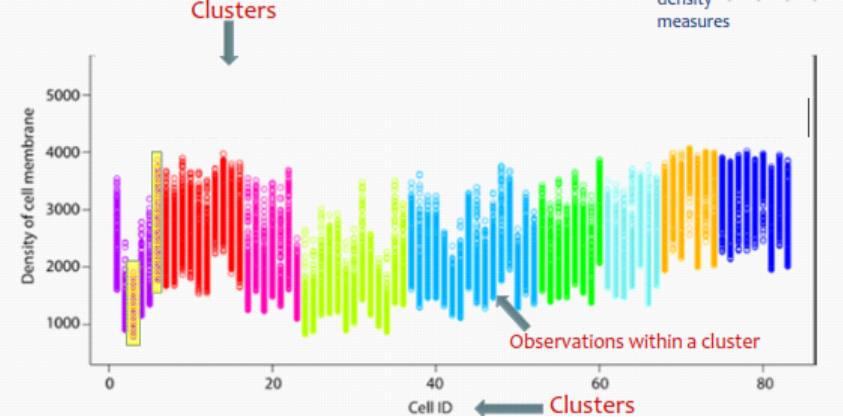
Dictated by your hypothesis/study design

T-test/ANOVA - Two-sided

Paired t-test - One-sided

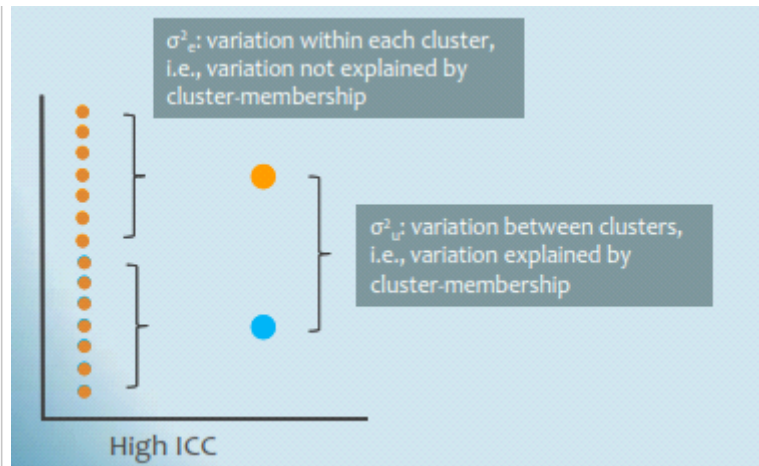
# 6 - Multilevel analysis, handling clustered data

<p><b>What is the difference between longitudinal data and clustered data?</b></p>	<p><b>Multilevel analysis</b></p> <p>General assumption of statistics: Data are a random sample from the population</p> <ol style="list-style-type: none"> <li>1. Longitudinal data - Dependency over time (paired t-tests)</li> <li>2. Clustered data - Data are not randomly sampled</li> </ol>
<p><b>Why the violation of assumption of independence creates more false positives?</b></p>	<p><b>Violating the assumption of independence</b></p> <p>Within the class - Scores are more alike (they are clustered), dependent by design</p> <p>This creates a high false positive rate</p>
<p><b>What is hierarquical data?</b></p>	<p><b>Conceptual introduction</b></p> <p>Nested/clustered/hierarquical data - Multiple measurements of the same object</p> <ul style="list-style-type: none"> <li>- Children in the same class (child is level 1, class is level 2, school is level 3)</li> </ul>  <p>Classical example: children in classes in schools</p> <ul style="list-style-type: none"> <li>- Neurons in the same pup, multiple axons/dendrites from the same neuron</li> <li>- Patients from the same doctor</li> </ul> <p>53% of neuroscience data in papers included nested data</p>
<p><b>What are the two types of nested designs?</b></p>	<p><b>Two types of nested designs</b></p> <ol style="list-style-type: none"> <li>1) All observations from the same cluster are in the same experimental condition</li> </ol>  <ol style="list-style-type: none"> <li>2) Observation from the same cluster can be part of different experimental conditions</li> </ol> <p>Research design II</p>

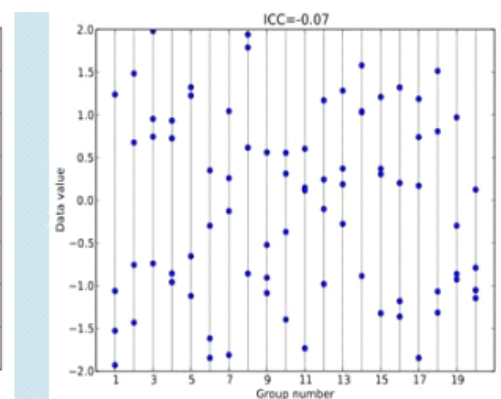
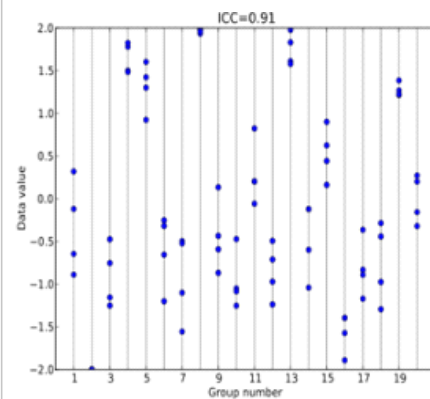
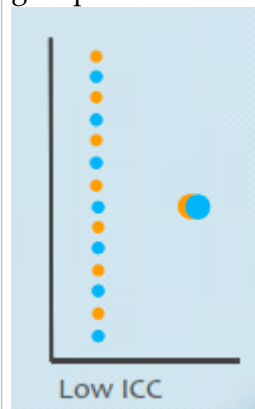
	
<p><b>What is the difference between observed and effective sample size?</b></p>	<p>Observed vs effective sample size  3 WT pups: 10 neurons per group  3 KO pups: 10 neurons per group</p> <p>Observed sample size: 60 measures  Effective sample size: Ranges between 6-60 depends on ICC</p> <p>ICC: Intra-cluster correlation  0 - All samples are independent  1 - All samples convey the same information</p>
<p><b>How can clusters be visually identified in a histogram?</b></p>	<p>Density of cell membrane - Observations of the same cell are more similar to each other</p>  <p>You cannot do ANOVA/t-tests on hierarchical data</p>
<p><b>What does ICC stand for?</b>  <b>What is the formula for ICC?</b></p>	<p>ICC - Ratio of variation in the data explained by clustering/Total variation in the data</p> $ICC = \frac{\sigma_u^2}{\sigma_u^2 + \sigma_e^2}$
<p><b>Visually, what is the difference between high</b></p>	<p><b>High ICC</b> - Small variation between groups, large variation among groups</p>



and low ICC?



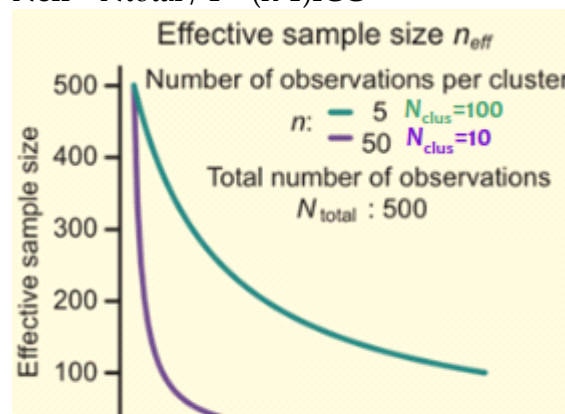
Low ICC - High variation between groups, small variation among groups



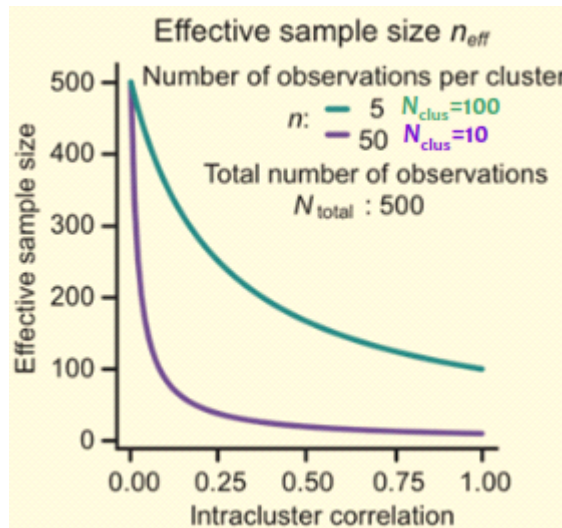
If the ICC is small, does that mean that you can perform a normal t-test/ANOVA with your data?

Effective sample size

$$N_{eff} = N_{total} / (1 + (n-1)ICC)$$



perform a normal t-test/ANOVA with your data?

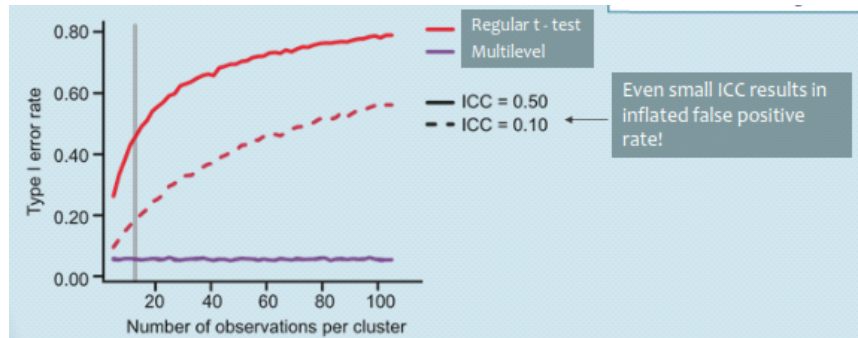


Even if ICC is small, it can have a large influence if your sample size is very large

What happens when you do ANOVA/t-test with clustered data with:

- a) Standard error
- b) Degrees of freedom
- c) p-values

If you do not do multilevel analysis with clustered data:  
Underestimate SE and p-value  
Overestimate t and df



Test is too significant - Type I error

What are the three ways that you can handle dependency in your data?

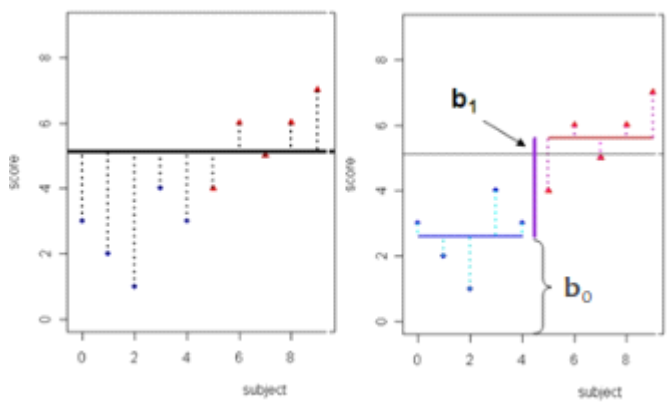
**Handling dependency**

- Average within each cluster - Lose information
- Take a random observation from each cluster - Lose even more information
- Multilevel analysis - Does not lose information, power of test increases

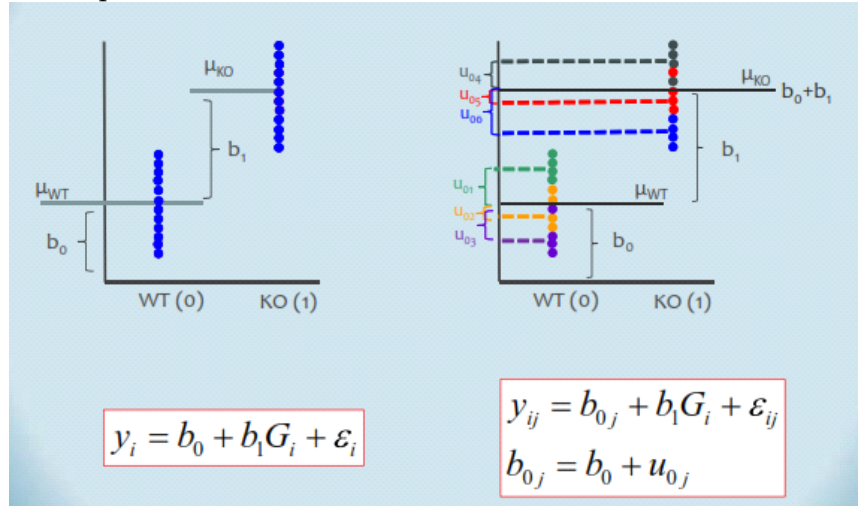
What is the only difference between normal regression and multilevel analysis?

**Normal regression**

- Intercept -  $b_0$
- Slope -  $b_1$



Only difference for Multilevel analysis: Submodel for the intercept! (Mean + deviation)



Cluster specific deviation - From the group mean

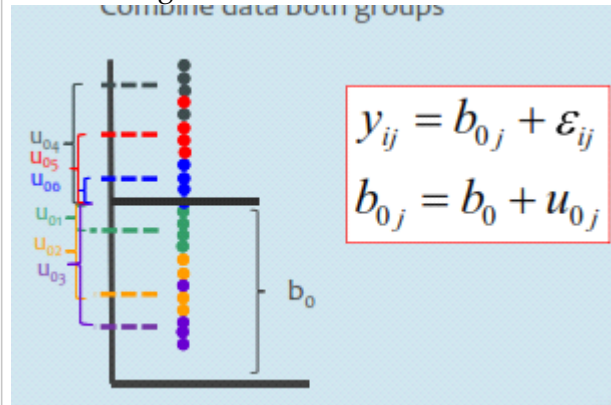
Sigma<sub>u</sub> - Intercept variance

Sigma<sub>ε</sub> - Within the clusters

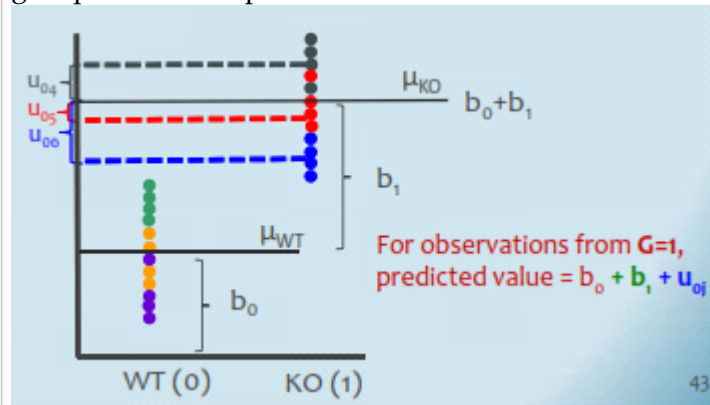
**What is the difference between intercept only model and multilevel models?**

Multilevel: Random intercept

**Overall ICC (intercept only model)**- Total variability in the data due to clustering



**Explained ICC (multilevel model)**- Variability present when taking group membership into account

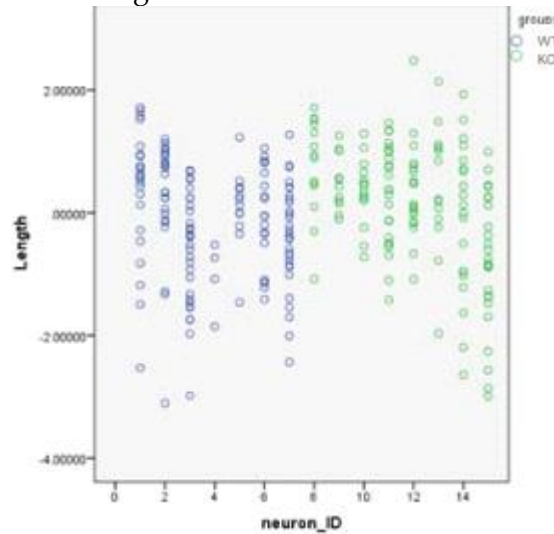


**Unexplained ICC** - Variability that remains after inclusion of group membership

What are the two different types of research design in multilevel analysis?

### Examples of research design

Dendrite length in WT and KO  
Clustering is observed in the data



Covariance is significant (difference between the clusters)-  
Multilevel model is necessary

No significant difference in length between genotype groups

Parameter	Estimate	Std. Error	df	t	Sig.	95% Confidence Interval	
						Lower Bound	Upper Bound
int	.025935	.115299	14.710	.225	.825	-.220240	.272111
ZGenotype	1.66198	.115468	14.712	1.439	.171	-.080336	.412733

a. Dependent Variable: Length.

Significant variation in mean dendrite length between neurons

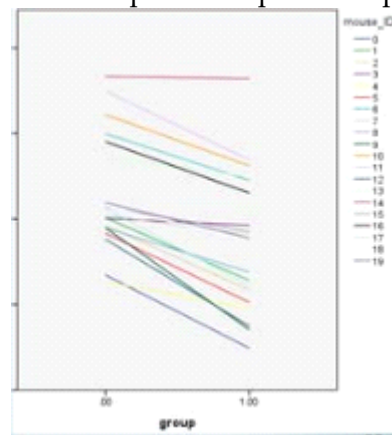
Parameter	Estimate	Std. Error	Wald Z	Sig.	95% Confidence Interval	
					Lower Bound	Upper Bound
Residual	.826775	.070595	11.712	.000	.699370	.977389
int (subject= neuron_ID) Variance	.148274	.071707	2.068	.039	.057466	.382575

a. Dependent Variable: Length.

### Research design II

Multiple observation from the same mouse, part is treated with virus, other part does not have virus

Cluster-specific intercept - Mean  
Cluster-specific slope - Manipulation effect



Fixed effect - Same for all clusters  
Random effect - Different for different clusters

	<p>All neurons are stained with GFP - part is transfected with virus, part is not</p> <ul style="list-style-type: none"> <li>○ Significant intercept variance</li> <li>○ Significant slope variance</li> </ul> <p><b>Always do multilevel analysis with clustered data, even if covariance parameters are not significant</b></p>
	<p><b>Conclusions</b></p> <p>Clustering needs to be taken into account in statistical test ML is just an extension regression analysis Read more about multilevel modeling</p>
	<p><b>Questions:</b></p> <ul style="list-style-type: none"> <li>• What is meant by “dependent data”</li> <li>• Observed versus effective sample size</li> <li>• What is the ICC</li> <li>• Why is multi-level modelling required if data are not independent</li> <li>• What is meant by “clustering” or “hierarchical data”?</li> </ul>

#### Assignment discussion

##### 5.1. Is sex still a predictor of ICV once height is taken into account?

Null hypothesis - Sex does not improve the prediction once height is taken into account

Alternative hypothesis - Sex does improve the prediction once height is taken into account

Check assumptions

Mean + standard deviation of groups need to be reported separately

The fact that height is not significant does not mean that you can exclude it from the model - you need to run the regression again without the variable

##### 5.2. Sepal length can be predicted by petal length/width

###### 1. Dummy coding

Two groups =  $a + bG$

-  $bG = 0$ ; intercept straight line

-  $A = 0$ ; sloped line  $b$

More than two groups =  $a + bG_1 + bG_2$

- Dummy coding - Compare each variable to the reference group (does not matter which one)

- Allows an ANOVA to be performed in the context of a regression

-  $B_1$  - Group 1 from reference group

-  $B_2$  - Group 2 from reference group

###### 2. Mislabelling

###### 3. Check for outliers

Petal length and petal width are highly correlated - Cannot both be in the model ( $VIF > 18$ )

You must include all dummies or no dummies - you cannot include only some dummies

Hierarchical entry - You know the predictors (from literature)

Stepwise entry - You do not know, you let the computer decide the best model

Two predictors - Petal length + species (do not count dummy codings individually)

# 6b. Power

What is the definition of Power?

**Power: Probability to detect an effect that is actually there**

$$\text{Power} = 1 - \beta$$

Probability of rejecting the null hypothesis if it is false

Possible research outcomes		H0 is	
		True	False
Researcher	Accept H0	True negative $1 - \alpha$	False negative Type II error ( $\beta$ )
	Reject H0	False positive Type I error ( $\alpha$ )	True positive $1 - \beta$ : power!

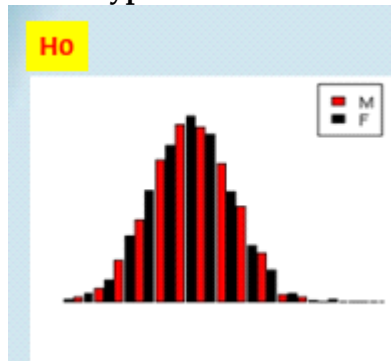
Why is there no type II error under the null hypothesis?

Example - Men and women differ in IQ scores

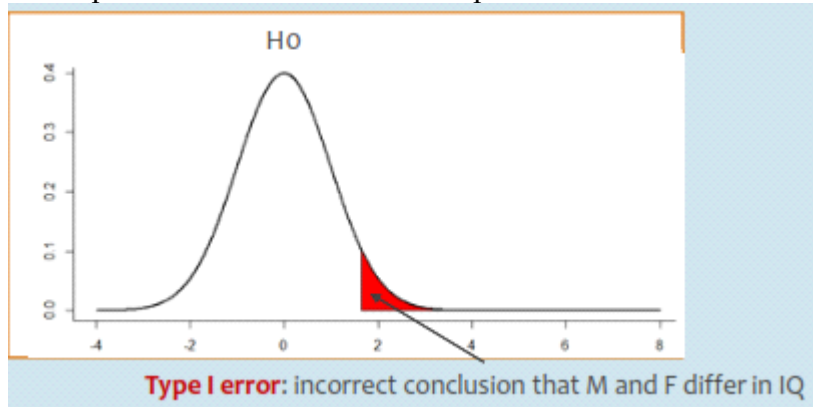
Sampling theory - Most results will be 0, but some test statistics will yield a false positive

How is type I, type II error and power represented in a t-statistic graph?

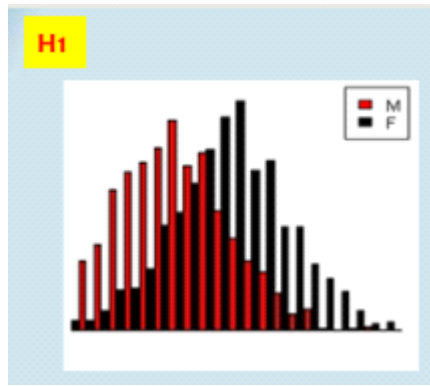
Under null hypothesis



Tail represents the chance of a false positive



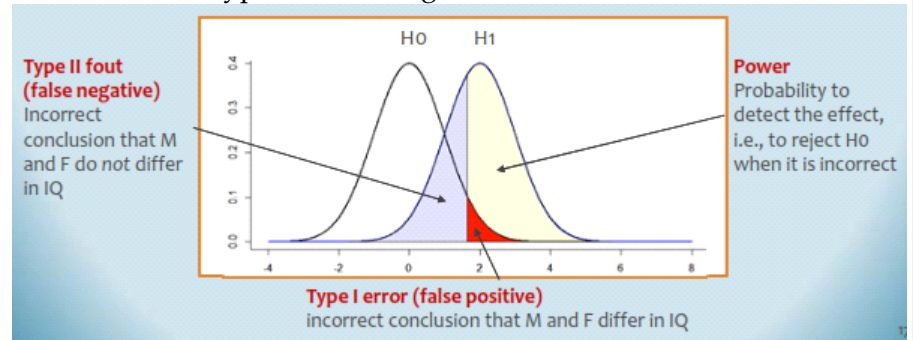
Under alternative hypothesis



Type I error - Right tail of left curve

Type II error - Left side of right curve

Power - Above type I error in rightmost curve



This is the distribution of t-values (test statistics - not the distribution of the data itself!)

**What is the relation between type I and type II errors?**

There is an inverse relation between Type I and Type II errors  
If alpha gets smaller, beta gets larger

**What is sensitivity and specificity?**

**Sensitivity** - Of all positives, how many are true positives  
Can you identify the sick  
**Specificity** - Of all negatives, how many are true negatives  
Can you identify the healthy

**What are the three ways you can increase power?**

**How to increase power?**

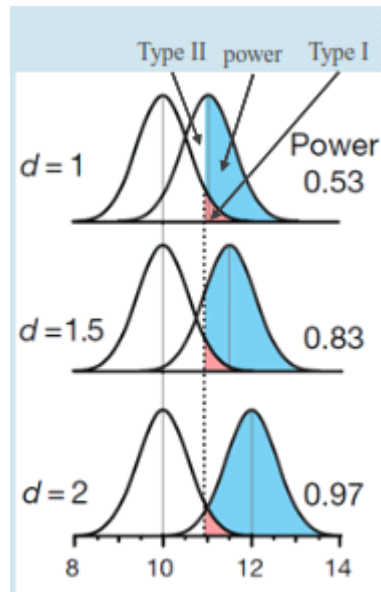
- Increasing sample size
- Increasing difference between two means (effect size)
- Decreasing variance

**How is the increase of effect size represented visually?**

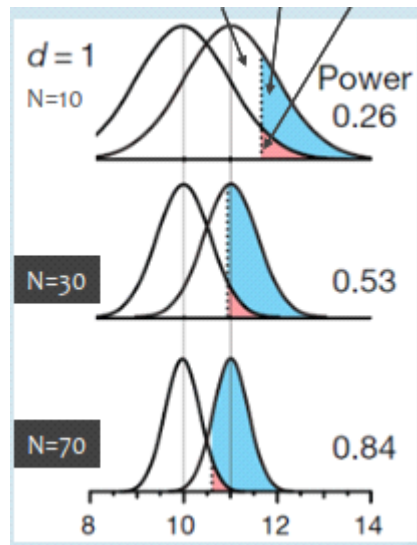
**Increase effect size** - Make test more difficult



How is the increase of sample size represented visually?



**Increase sample size** - Less sampling variation, more accurate estimate of the effect

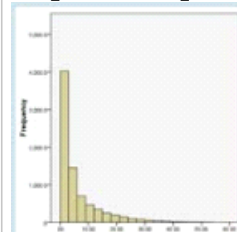


The curve becomes narrower (less variation)

Decreasing noise - Control through careful matching and selection; use of a good instrument (standardized, reliable, objective, high resolution)

How is power related to the way you collect your data?

**Depression questionnaire**

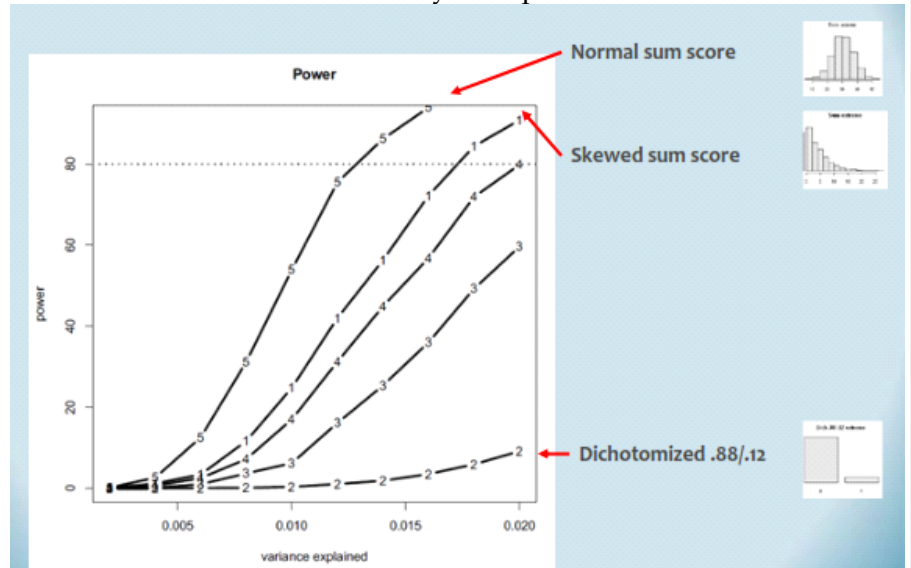


	Rarely /never	Sometimes	Occasionally	Often/always
6) I felt depressed	0	1	2	3
9) I thought my life had been a failure	0	1	2	3
10) I felt fearful	0	1	2	3
14) I felt lonely	0	1	2	3
17) I had crying spells	0	1	2	3
18) I felt sad	0	1	2	3
19) I felt that people disliked me	0	1	2	3

Extreme questions - Most people will score 0, even though they do not feel equally well

Controls = 0  
 Loss of a lot of information  
 Cases = 1

Dichotomized data has very little power



The instrument we use and statistical treatments change the power of your data

**What is censoring?**  
**What are `bottom` and `ceiling` effects?**

**Censoring** - Loss of information in your data  
**Bottom effect** - Some could not perform the task at all  
**Ceiling effect** - Some answered everything correctly

**What is the mean power of neuroscience studies?**  
**Why is that a problem?**

Current state of affairs  
 Neuroscience studies - **Mean power of 21%**  
 Power should be at least 80%

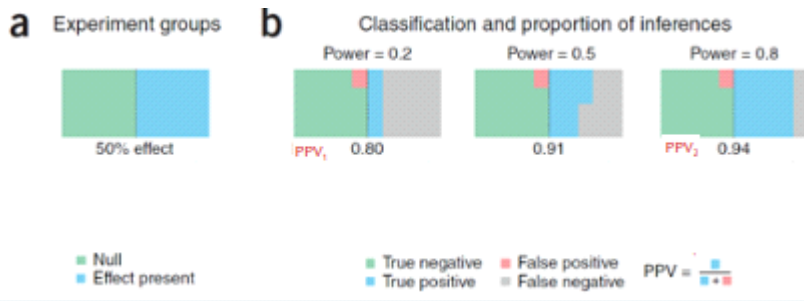
You cannot trust what is discovered  
 You cannot trust what is not discovered  
 Underpower studies are useless

**What is the problem with increasing false negatives due to underpowered studies?**

Problems with underpowered studies  
**1. More false negatives**  
 a. Most research focus on minimizing false positives  
 b. False negatives - Waste of money, effort and time; negative results cannot be interpreted as the absence of an effect

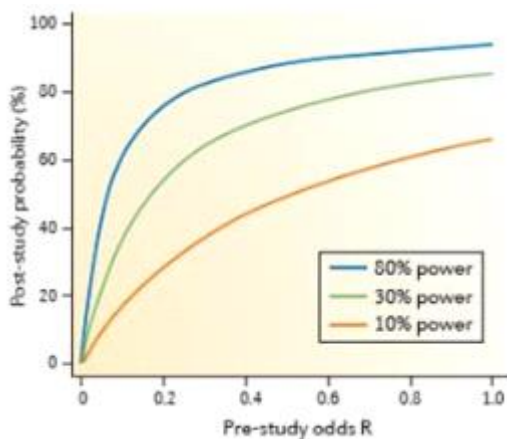
**What is PPV?**  
**What is the relation between PPV and power?**

**2. Observed effects are less likely to be true**  
 a. Positive predictive value - The effect that we observed is likely to be a true effect (depends on the knowledge of the field)  
 Shoe sizes varies between men and women - Safe hypothesis, needs very little evidence  
 Mind-reading - Risky study, it takes a lot of evidence to convince readers



Calculation of  $PPV = (1-\beta) * odds / ((1-\beta) * odds + \alpha)$   
 Odds = successes/failures  
 True positives / true + false positives

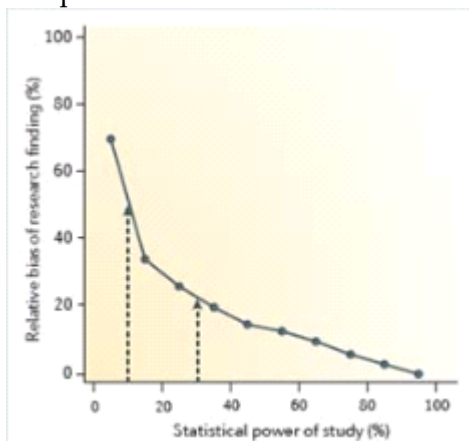
The probability that something is true determines the likelihood of what you will find is true



What is Winner's curse?  
 What is its relationship with power?

**3. Overestimation effects**

Winner's curse - Detected effects are overestimated/inflated  
 Large effect in false positives are more likely to occur with a small sample



Study replication often finds contradictory results

Why are underpowered studies unethical?

**4. Unethical**

Low powered studies are inefficient and wasteful  
 Sacrifice of animals, public funding wasted

Underpowered studies do not yield any conclusions

**Conclusion**

Given that I want to see an effect of at least a certain size - how can I design my study

A priori testing - G-power software

Small power - DO not do the study or increase sample size

**Questions**

- What is power
- False positives/false negatives (the table)
- Factors that affect power (e.g., N, effect size, variance: learn to interpret statistical equations (like t-test) in terms of power)
- Consequences of poor power (e.g., false negatives, Winner's curse, PPV, unethical)