## Data and variables

DATA: the answers to questions or measurements from the experiment

VARIABLE = measurement which varies between subjects e.g. height or gender


| 4 | A | B | c | 0 |
| :---: | :---: | :---: | :---: | :---: |
| 1 | Subject ID | Gender | Year of study | Height |
| 2 | 1 | Male | 1 | 170 |
| 3 | 2 | Female | 2 | 160 |
| 4 | 3 | Female | 3 | 165 |
| 5 | 4 | Male | PG | 175 |
| 6 | 5 | Female | 3 | 168 |

## Data types



## Data types



# Questionnaire for GCSE Maths Pupils What data types relate to following questions? 

> Q1: What is your favourite subject?
Maths $\quad$ English Science Art French

- Q2: Gender:


## Male Female

- Q3: I consider myself to be good at mathematics:

| Strongly <br> Disagree | Disagree | Not Sure | Agree | Strongly <br> Agree |
| :--- | :--- | :--- | :--- | :--- |

> Q4: Score in a recent mock GCSE maths exam:
Score between 0\% and 100\%

# Questionnaire for GCSE Maths Pupils What data types relate to following questions? 

- Q1: What is your favourite subject?
- Q2: Gender:

$$
\text { Male } \quad \text { Female }
$$

- Q3: I consider myself to be good at mathematics:

| Strongly <br> Disagree | Disagree | Not Sure | Agree | Strongly <br> Agree |
| :--- | :--- | :--- | :--- | :--- |

Ordinal
$>$ Q4: Score in a recent mock GCSE maths exam:
Score between 0\% and 100\%
Scale

## Populations and samples

- Taking a sample from a population


Sample data 'represents' the whole population

## Point estimation

Sample data is used to estimate parameters of a population

Statistics are calculated using sample data.
Parameters are the characteristics of population data


## How can exam score data be summarised?

Exam marks for 60 students (marked out of 65)

| 48 | 37 | 1 | 33 | 26 | 22 | 15 | 22 | 40 | 30 | 12 | 36 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 21 | 20 | 29 | 13 | 44 | 52 | 28 | 39 | 16 | 48 | 56 | 27 |
| 47 | 12 | 35 | 24 | 10 | 36 | 18 | 34 | 9 | 25 | 31 | 42 |
| 31 | 27 | 64 | 25 | 58 | 17 | 26 | 38 | 28 | 43 | 33 | 5 |
| 25 | 55 | 7 | 32 | 39 | 23 | 49 | 43 | 11 | 37 | 22 | 54 |

## Summary statistics

- Mean $=\quad \frac{\sum_{i=1}^{n} x}{n}=\bar{x}$

Standard deviation (s) is a measure of how much the individuals differ from the mean

$$
s=\sqrt{\frac{\sum_{i=1}^{n}\left(x_{i}-\bar{x}\right)^{2}}{n-1}}
$$

Large SD = very spread out data
Small $S D=$ there is little variation from the mean
For exam scores, mean $=30.5, S D=14.46$

## Interpretation of standard deviation

- The larger the standard deviation, the more spread out the data is.



## Discussion

- How could you explain to a student what we mean by data being assumed to follow a Normal Distribution?


## Group Frequency Table

|  | Frequency | Percent |
| :---: | :---: | :---: |
| 0 but less than 10 | 4 | 6.7 |
| 10 but less than 20 | 9 | 15.0 |
| 20 but less than 30 | 17 | 28.3 |
| 30 but less than 40 | 15 | 25.0 |
| 40 but less than 50 | 9 | 15.0 |
| 50 but less than 60 | 5 | 8.3 |
| 60 or over | 1 | 1.7 |
| Total | $\mathbf{6 0}$ | $\mathbf{1 0 0 . 0}$ |

## Histogram and Probability Distribution for Exam Marks Data



## Histogram and Probability Distribution for Exam Marks Data



## IQ is normally distributed



Mean $=100, S D=15.3$

## $95 \% 1.96 \times$ SD's from the mean



$$
\begin{aligned}
& \text { mean }-(1.96 \times S D) \\
& 100-(1.96 \times 15.3)=70
\end{aligned}
$$

$$
\begin{aligned}
& \text { mean }+(1.96 \times S D) \\
& 100+(1.96 \times 15.3)=130
\end{aligned}
$$

$95 \%$ of people have an IQ between 70 and 130

## Assessing Normality

Charts can be used to informally assess whether data is:

Normally distributed

## Or....Skewed




The mean and median are very different for skewed data.

## Discussion

## Is the following statement:

" 2 out of 3 people earn less than the average income"
A. True
B. False
C. Don't know

## Sometimes the median makes more sense!

Chart 1.2 (BHC): Income distribution for the whole population, 2011/12
Median income $£ 427$ pw


Source: Households Below Average Income: An analysis of the income distribution 1994/95-2011/12, Department for Work and Pensions

## Choosing summary statistics



## Which graph? Exercise

| $1^{\text {st }}$ variable | Only 1 variable | Scale | Categorical |
| :--- | :--- | :--- | :--- |
| Scale | Histogram | Scatter plot | Box-plot/ <br> Confidence <br> interval plot |
| Categorical | Pie/ Bar | Box-plot/ <br> Confidence <br> interval plot | Stacked/ <br> multiple bar <br> chart |

Which graph would you use when investigating:

1) Whether daily temperature and ice cream sales were related?
2) Comparison of mean reaction time for a group having alcohol and a group drinking water

## Exercise: Ticket cost comparison

Summary statistics for cost of Titanic ticket by survival

| Cost of ticket | Survived? |  |
| :--- | :---: | :---: |
|  | Died | Survived |
| Mean | 23.4 | 49.4 |
| Median | 10.5 | 26 |
| Standard Deviation | 34.2 | 68.7 |
| Interquartile range | 18.2 | 46.6 |
| Minimum | 0 | 0 |
| Maximum | 263 | 512.33 |

a) Is there a big difference in average ticket price by group?
b) Which group has data which is more spread out?
c) Is the data skewed?
d) Is the mean or median a better summary measure?

## Which graph? Solution

| $1^{\text {st }}$ variable | Only 1 variable | Scale | Categorical |
| :--- | :--- | :--- | :--- |
| Scale | Histogram | Scatter plot | Box-plot/ <br> Confidence <br> interval plot |
| Categorical | Pie/ Bar | Box-plot/ <br> Confidence <br> interval plot | Stacked/ <br> multiple bar <br> chart |

Which graph would you use when investigating:

1) Whether daily temperature and ice cream sales were related? Scatter
2) Comparison of mean reaction time for a group having alcohol and a group drinking water Boxplot or confidence interval plot

## Exercise: Ticket cost comparison Solution

a) Is there a big difference in average ticket price by group?
The mean and median are much bigger in those who survived.
b) Which group has data which is more spread out?

The standard deviation and interquartile range are much bigger for those who survived so that data is more spread out
c) Is the data skewed?

Yes. The medians are much smaller than the means and the plots show the data is positively skewed.
d) Is the mean or median a better summary measure?

The median as the data is skewed

## Comparing means

## Summarising means

## Calculate summary statistics by group

## - Look for outliers/ errors

- Use a box-plot or confidence interval plot

Box-plot comparing housework per week by gender


# T-tests <br> <br> Paired or Independent (Unpaired) Data? 

 <br> <br> Paired or Independent (Unpaired) Data?}

T-tests are used to compare two population means

- Paired data: same individuals studied at two different times or under two conditions PAIRED T-TEST
- Independent: data collected from two separate groups INDEPENDENT SAMPLES T-TEST


## Comparison of hours worked in 1988 to today

Paired or unpaired?
If the same people have reported their hours for 1988 and 2014 have PAIRED measurements of the same variable (hours)
Paired Null hypothesis: The mean of the paired differences $=0$

$$
H_{0}: \mu_{d}=0
$$

If different people are used in 1988 and 2014 have independent measurements
Independent Null hypothesis: The mean hours worked in 1988 is equal to the mean for 2014

$$
H_{0}: \mu_{1988}=\mu_{2014}
$$

## SPSS data entry

## Paired Data

|  |  |  |  |
| :--- | :--- | ---: | ---: |
|  | Name | Hours1988 | Hours2014 |
| 1 | Joe Bloggs | 35 | 38 |
| 2 | Sam Smith | 37 | 35 |
| 3 | Joyce Jones | 20 | 35 |
| 4 |  |  |  |

## Independent Groups

|  | Name | Hours | Year |
| :--- | :--- | ---: | ---: |
| 1 | Joe Bloggs | 35 | 1988 |
| 2 | Sam Smith | 37 | 1988 |
| 3 | Joyce Jones | 20 | 1988 |
| 4 | Li Yu | 38 | 2014 |
| 5 | Sally McGregor | 35 | 2014 |
| 6 | Balvinder Singh | 35 | 2014 |
| 7 |  |  |  |

## What is the t-distribution?

- The t -distribution is similar to the standard normal distribution but has an additional parameter called degrees of freedom (df or v)

For a paired t -test, $\mathrm{v}=$ number of pairs - 1
For an independent t-test, $\quad v=n_{\text {group } 1}+n_{\text {group } 2}-2$

- Used for small samples and when the population standard deviation is not known
- Small sample sizes have heavier tails


## Relationship to normal

- As the sample size gets big, the t -distribution matches the normal distribution



## CAST e-books in statistics: t-distribution

http://cast.massey.ac.nz/collection_public.html

- Introductory e-book (general) and select 10. Testing Hypotheses

3. Tests about means
4. The $t$-distribution
5. The t-test for a mean

## Exercise

- For Examples 1 and 2 (on the following four slides) discuss the answers to the following:
- State a suitable null hypothesis
- State whether it's a Paired or Independent Samples t-test
- Decide whether to reject the null hypothesis
- State a conclusion in words


## Example 1: Triglycerides

- In a weight loss study, Triglyceride levels were measured at baseline and again after 8 weeks of taking a new weight loss treatment.



## Example 1: t-Test Results

|  |  |  |  |  |  | t | df | Sig. (2tailed) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Mean | Std. <br> Deviation | Std. Error Mean | 95\% Confidence Interval of the Difference |  |  |  |  |
|  |  |  |  | Lower | Upper |  |  |  |
| Triglyceride level at week 8 (mg/dl) Triglyceride level at baseline (mg/dl) | -11.371 | 80.360 | 13.583 | -38.976 | 16.233 | -. 837 | 34 | . 408 |

## Null Hypothesis is:

P -value $=$
Decision (circle correct answer): Reject Null/ Do not reject Null

Conclusion:

## Example 1: Solution

|  |  |  |  |  |  | t | df | Sig. (2tailed) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Mean | Std. Deviation | Std. Error Mean | 95\% Confidence Interval of the Difference |  |  |  |  |
|  |  |  |  | Lower | Upper |  |  |  |
| Triglyceride level at week 8 (mg/dl) - Triglyceride level at baseline (mg/dl) | -11.371 | 80.360 | 13.583 | -38.976 | 16.233 | -. 837 | 34 | . 408 |

$$
H_{0}: \mu_{d}=0
$$

As p $>0.05$, do NOT reject the null

NO evidence of a difference in the mean triglyceride before and after treatment


## Example 2: Weight Loss

- Weight loss was measured after taking either a new weight loss treatment or placebo for 8 weeks

| Treatmen <br> t group | N | Mean | Std. <br> Deviation |
| :--- | :---: | :---: | :---: |
| Placebo | 19 | -1.36 | 2.148 |
| New drug | 18 | -5.01 | 2.722 |



|  | Leverie's Test or Equality of Variances |  | T-test results |  |  |  |  | $95 \% \mathrm{Cl}$ of the Difference |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | F | Sig. | t | df | Sig. (2-tailed) | Mean Difference | Std. Error Difference | Lower | Upper |
| Equal variances assumed | 2.328 | . 136 | 4.539 | 35 | . 000 | 3.648 | . 804 | 2.016 | 5.280 |
| Equal variances not assumed |  |  | 4.510 | 32.342 | . 000 | 3.648 | . 809 | 2.001 | 5.295 |

Null Hypothesis is:

## P-value = <br> Decision (circle correct answer): Reject Null/ Do not reject Null

## Conclusion:

|  | Leverie's Test ior Equality of Variances |  | T-test results |  |  |  |  | $95 \% \mathrm{Cl}$ of the Difference |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | F | Sig. | t | df | $\begin{array}{c\|} \hline \text { Sig. } \\ \text { (2-tailed) } \end{array}$ | Mean Difference | Std. Error Difference | Lower | Upper |
| Equal variances assumed | 2.328 | . 136 | 4.539 | 35 | . 000 | 3.648 | 804 | 2.016 | 5.280 |
| Equal variances not assumed |  |  | 4.510 | 32.342 | . 000 | 3.648 | . 809 | 2.001 | 5.295 |

$$
\mathrm{H}_{0}: \mu_{\text {new }}=\mu_{\text {placebo }} \quad \text { As } \mathrm{p}<0.05, \text { DO reject the null }
$$

## IS evidence of a

 difference in weight loss between treatment and placebo

## Assumptions

- Every test has assumptions
- Tutors quick guide shows assumptions for each test and what to do if those assumptions are not met


## Assumptions in t-Tests

- Normality: Plot histograms
- One plot of the paired differences for any paired data
- Two (One for each group) for independent samples
- Don't have to be perfect, just roughly symmetric
- Equal Population variances: Compare sample standard deviations
- As a rough estimate, one should be no more than twice the other
- Do an F-test (Levene's in SPSS) to formally test for differences
- However the $t$-test is very robust to violations of the assumptions of Normality and equal variances, particularly for moderate (i.e. $>30$ ) and larger sample sizes


## Histograms from Examples 1 and 2



Do these histograms look approximately normally distributed?

## Levene's Test for Equal Variances from Examples 2

|  | Levene's Test <br> for Equality of <br> Variances |  | T-test results |  |  |  |  |  |  | $95 \%$ Cl of the <br> Difference |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | F | Sig. | t | df | Sig. <br> $(2$-tailed) | Mean <br> Difference | Std. Error <br> Difference | Lower | Upper |  |  |
| Equal variances <br> assumed | 2.328 | .136 | 4.539 | 35 | .000 | 3.648 | .804 | 2.016 | 5.280 |  |  |
| Equal variances <br> not assumed |  |  | 4.510 | 32.342 | .000 | 3.648 | .809 | 2.001 | 5.295 |  |  |

Null hypothesis is that pop variances are equal
i.e. $H_{0}: \sigma_{\text {new }}^{2}=\sigma^{2}{ }_{\text {placebo }}$

Since $p=0.136$ and so is $>0.05$ we do not reject the null i.e. we can assume equal variances ©

## What if the assumptions are not met?

- There are alternative tests which do not have these assumptions

| Test | Check | Equivalent <br> non-parametric test |
| :--- | :--- | :--- |
| Independent t-test | Histograms of data <br> by group | Mann-Whitney |
| Paired t-test | Histogram of paired <br> differences | Wilcoxon signed rank |

## Sampling Variation

Every sample taken from a population, will contain different numbers so the mean varies.

Which estimate is most reliable?

How certain or uncertain are we?

## Confidence Intervals

- A range of values within which we are confident (in terms of probability) that the true value of a pop parameter lies
- A $95 \% \mathrm{Cl}$ is interpreted as $95 \%$ of the time the Cl would contain the true value of the pop parameter
- i.e. $5 \%$ of the time the CI would fail to contain the true value of the pop parameter

|  |  |  |  |  |  | t | df | Sig. (2tailed) |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Mean | Std. Deviation | Std. Error Mean | 95\% Confidence Interval of the Difference |  |  |  |  |
|  |  |  |  | Lower | Upper |  |  |  |
| Triglyceride level at week 8 (mg/dl) Triglyceride level at baseline (mg/dl) | -11.371 | 80.360 | 13.583 | -38.976 | 16.233 | -. 837 | 34 | . 408 |

## CAST e-books in statistics: Confidence Intervals

http://cast.massey.ac.nz/collection_public.html

Choose introductory e-book (general) and select
9. Estimating Parameters
3. Conf. Interval for mean
6. Properties of 95\% C.I.

## Confidence Interval simulation from CAST



Seven do not include 141.1 mmHg - we would expect that the $95 \%$ Cl will not include the true population mean $5 \%$ of the time

## Exercise

- Discuss what the interpretation is for the confidence interval from Example 2 (Weight loss was measured after taking either a new weight loss treatment or placebo for 8 weeks) highlighted below:

|  | Levene's Test for Equality of Variances |  | T-test results |  |  |  |  | $95 \% \mathrm{Cl}$ of the Difference |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | F | Sig. | t | df | $\begin{gathered} \text { Sig. } \\ \text { (2-tailed) } \end{gathered}$ | Mean Difference | Std. Error Difference | Lower | Upper |
| Equal variances assumed | 2.328 | . 136 | 4.539 | 35 | . 000 | 3.648 | . 804 | 2.016 | 5.280 |

## Exercise: Solution

- Discuss what the interpretation is for the confidence interval from Example 2 highlighted below:

|  | Levene's Test for Equality of Variances |  | T-test results |  |  |  |  | $95 \% \mathrm{Cl}$ of the Difference |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | F | Sig. | t | df | Sig. (2-tailed) | Mean Difference | Std. Error Difference | Lower | Upper |
| Equal variances assumed | 2.328 | . 136 | 4.539 | 35 | . 000 | 3.648 | . 804 | 2.016 | 5.280 |

The true mean weight loss would be between about 2 to 5 kg with the new treatment.
This is always positive hence the hypothesis test rejected the null that the difference is zero

## Choosing the right test

## Choosing the right test

- One of the most common queries in stats support is 'Which analysis should I use'
- There are several steps to help the student decide
- When a student is explaining their project, these are the questions you need answers for


## Choosing the right test

1) A clearly defined research question
2) What is the dependent variable and what type of variable is it?
3) How many independent variables are there and what data types are they?
4) Are you interested in comparing means or investigating relationships?
5) Do you have repeated measurements of the same variable for each subject?

## Research question

Clear questions with measurable quantities

- Which variables will help answer these questions
- Think about what test is needed before carrying out a study so that the right type of variables are collected


## Dependent variables



Does attendance have an association with exam score?

Do women do more housework than men?

## What variable type is the dependent?



## Are boys better at maths?

- How can 'better' be measured and what type of variable is it?
Exam score (Scale)
- Do boys think they are better at maths??

I consider myself to be good at maths (ordinal)

## How many variables are involved?

- Two - interested in the relationship
- One dependent and one independent
- One dependent and several independent variables: some may be controls
- Relationships between more than two: multivariate techniques (not covered here)


## Data types

| Research question | Dependent/ <br> outcome variable | Independent/ <br> explanatory variable |
| :--- | :--- | :--- |
| Does attendance have an <br> association with exam score? | Exam score (scale) | Attendance (Scale) |
| Do women do more <br> housework than men? | Hours of <br> housework per <br> week (Scale) | Gender (binary) |

## Exercise:

How would you investigate the following topics? State the dependent and independent variables and their variable types.

| Research question | Dependent/ <br> outcome variable | Independent/ <br> explanatory variable |
| :--- | :--- | :--- |
| Were Americans more likely to <br> survive on board the Titanic? |  |  |
| Does weekly hours of work <br> influence the amount of time <br> spent on housework? |  |  |
| Which of 3 diets is best for <br> losing weight? |  |  |

## Exercise: Solution

 How would you investigate the following topics? State the dependent and independent variables and their variable types.| Research question | Dependent/ <br> outcome variable | Independent/ <br> explanatory variable |
| :--- | :--- | :--- |
| Were Americans more likely to <br> survive on board the Titanic? | Survival (Binary) | Nationality (Nominal) |
| Does weekly hours of work <br> influence the amount of time <br> spent on housework? | Hours of housework <br> (Scale) | Hours of work (Scale) |
| Which of 3 diets is best for <br> losing weight? | Weight lost on diet <br> (Scale) | Diet (Nominal) |

## Comparing means

- Dependent = Scale
- Independent = Categorical
- How many means are you comparing?
- Do you have independent groups or repeated measurements on each person?


## Comparing measurements on the same people

Also known as within group comparisons or repeated measures.

Can be used to look at differences in mean score:
(1) over 2 or more time points e.g. 1988 vs 2014
(2) under 2 or more conditions e.g. taste scores

Participants are asked to
taste 2 types of cola and give
each scores out of 100.
Dependent = taste score
Independent = type of cola

## Comparing means



## Comparing means



## Exercise - Comparing means

| Research question | Dependent <br> variable | Independent <br> variable | Test |
| :--- | :--- | :--- | :--- |
| Do women do more <br> housework than men? | Housework (hrs <br> per week) <br> (Scale) | Gender <br> (Nominal) |  |
| Does Margarine X <br> reduce cholesterol? <br> Everyone has <br> cholesterol measured <br> on 3 occasions | Cholesterol <br> (Scale) | Occasion <br> (Nominal) |  |
| Which of 3 diets is best <br> for losing weight? | Weight lost on <br> diet (Scale) | Diet <br> (Nominal) |  |

## Exercise: Solution

| Research question | Dependent <br> variable | Independent <br> variable | Test |
| :--- | :--- | :--- | :--- |
| Do women do more <br> housework than men? | Housework (hrs <br> per week) <br> (Scale) | Gender <br> (Nominal) | Independent t- <br> test |
| Does Margarine X <br> reduce cholesterol? <br> Everyone has <br> cholesterol measured <br> on 3 occasions | Cholesterol <br> (Scale) | Occasion <br> (Nominal) | Repeated <br> measures ANOVA |
| Which of 3 diets is best <br> for losing weight? | Weight lost on <br> diet (Scale) | Diet <br> (Nominal) | One-way ANOVA |

## Tests investigating relationships

| Investigating <br> relationships between | Dependent <br> variable | Independent <br> variable | Test |
| :--- | :--- | :--- | :--- |
| 2 categorical variables | Categorical | Categorical | Chi-squared test |
| 2 Scale variables | Scale | Scale | Pearson's correlation |
| Predicting the value of <br> an dependent variable <br> from the value of a <br> independent variable | Scale | Scale/binary | Simple Linear <br> Regression |
|  | Binary | Scale/ binary | Logistic regression |

Note: Multiple linear regression is when there are several independent variables

## Exercise: Relationships

| Research question | Dependent <br> variable | Independent <br> variables | Test |
| :--- | :--- | :--- | :--- |
| Does attendance <br> affect exam score? | Exam score <br> (Scale) | Attendance <br> (Scale) |  |
| Do women do more <br> housework than men? | Housework <br> (hrs per week) <br> (scale) | Gender (Binary) <br> Hours worked <br> (Scale) |  |
| Were Americans more <br> likely to survive on <br> board the Titanic? | Survival <br> (Binary) | Nationality <br> (Nominal) |  |
|  | Survival <br> (Binary) | Nationality, <br> Gender, class |  |

Note: There may be 2 appropriate tests for some questions

## Exercise: Solution

| Research question | Dependent <br> variable | Independent <br> variables | Test |
| :--- | :--- | :--- | :--- |
| Does attendance <br> affect exam score? | Exam score <br> (Scale) | Attendance <br> (Scale) | Correlation/ <br> regression |
| Do women do more <br> housework than men? | Housework <br> (hrs per week) <br> (scale) | Gender (Binary) <br> Hours worked <br> (Scale) | Regression |
| Were Americans more <br> likely to survive on <br> board the Titanic? | Survival <br> (Binary) | Nationality <br> (Nominal) | Chi-squared |
|  | Survival <br> (Binary) | Nationality, <br> Gender, class | Logistic <br> regression |

Note: There may be 2 appropriate tests for some questions

## Non-parametric tests

## Parametric or non-parametric?

## Statistical tests fall into two types:



Nonparametric
Non-parametric techniques are usually based on ranks/ signs rather than actual data

## Ranking raw data

- Nonparametric techniques are usually based on ranks or signs
- Scale data is ordered and ranked
- Analysis is carried out on the ranks rather than the actual data

|  | Original variable | Rank of subject |
| :---: | :---: | :---: |
| Group | ReactionTime | Rank |
| Placebo | - 37 | 1 |
| Placebo | - 38 | 2 |
| Placebo | . 61 | 3 |
| Placebo | . 78 | 4 |
| Placebo | . 83 | 5 |
| Placebo | . 86 | 6 |
| Placebo | - 90 | 7 |
| Placebo | . 95 | 8 |
| Alcohol | . 98 | 9 |
| Alcohol | 1.11 | 10 |
| Alcohol | 1.27 | 11 |
| Alcohol | 1.32 | 12 |
| Alcohol | 1.44 | 13 |
| Alcohol | 1.45 | 14 |
| Alcohol | 1.46 | 15 |
| Placebo | 1.63 | 16 |
| Alcohol | 1.76 | 17 |
| Placebo | 1.97 | 18 |
| Alcohol | 2.56 | 19 |
| Alcohol | 3.07 | 20 |

## Non-parametric tests

- Non-parametric methods are used when:
- Data is ordinal
- Data does not seem to follow any particular shape or distribution (e.g. Normal)
- Assumptions underlying parametric test not met
- A plot of the data appears to be very skewed
- There are potential influential outliers in the dataset
- Sample size is small

Note: Parametric tests are fairly robust to non-normality. Data has to be very skewed to be a problem


## Do these look normally distributed?



## Do these look normally distributed?



## What can be done about non-normality?

If the data are not normally distributed, there are two options:

1. Use a non-parametric test
2. Transform the dependent variable

Histogram of vbarg\$A1size

For positively skewed data, taking the log of the dependent variable often produces normally distributed values


Histogram of vbarg\$logA1


## Non-parametric tests

| Parametric test | What to check for <br> normality | Non-parametric test |
| :--- | :--- | :--- |
| Independent t-test | Dependent variable by <br> group | Mann-Whitney test |
| Paired t-test | Paired differences | Wilcoxon signed rank test |
| One-way ANOVA | Residuals/Dependent | Kruskal-Wallis test |
| Repeated measures <br> ANOVA | Residuals | Friedman test |
| Pearson's Correlation <br> Co-efficient | At least one of the <br> variables should be <br> normal | Spearman's Correlation <br> Co-efficient |
| Linear Regression | Residuals | None - transform the data |

Notes: The residuals are the differences between the observed and expected values.

## Exercise: Comparison of housework

Which test should be carried out to compare the hours of housework for males and females? Look at the histograms of housework by gender to decide.


## Exercise: Solution

Which test should be carried out to compare the hours of housework for males and females?

The male data is very skewed so use the Mann-Whitney.


## Summary



# Scenario - questions to consider Research question 

Dependent variable and type:

Independent variables:

Are there repeated measurements of the same variable for each subject?

## Friedman results

The student has never studied hypothesis testing before. Explain the concepts and what a p-value is. Is there a difference between the scores given to the preferences of the different criteria?

Hypothesis Test Summary

|  | Null Hypothesis | Test | Sig. |
| :---: | :---: | :---: | :---: |
| 1 | The distributions of Visual Feedback, Price Update and Virtual use of Product are the same. | Related- <br> Samples <br> Friedman's <br> Two-Way <br> Analysis of <br> Variance by <br> Ranks | . 358 |

SPSS: Analyze $\rightarrow$ Non-parametric Tests $\rightarrow$ Related Samples

## Resume

Do you feel that you

1. Have improved your ability to recall and describe some basic applied statistical methods?
2. Have developed ideas for explaining some of these methods to students in a maths support setting?
3. Feel more confident about providing statistics support with the topics covered?

## Mann Whitney Test

## Legal drink drive limits

Research question: Does drinking the legal limit for alcohol affect driving reaction times?

- Participants were given either alcoholic or nonalcoholic drinks and their driving reaction times tested on a simulated driving situations
They did not know whether they had the alcohol or not

```
Placebo: 0.90 0.37 1.63 0.83 0.95 0.78 0.86 0.61 0.38 1.97
Alcohol: 1.46 1.45 1.76 1.44 1.11 3.07 0.98 1.27 2.56 1.32
```


## Mann-Whitney test

- Nonparametric equivalent to independent t-test
- The data from both groups is ordered and ranked
- The mean rank for the groups is compared

| Group | ReactionTime | Rank |
| :--- | ---: | ---: |
| Placebo | .37 | 1 |
| Placebo | .38 | 2 |
| Placebo | .61 | 3 |
| Placebo | .78 | 4 |
| Placebo | .83 | 5 |
| Placebo | .86 | 6 |
| Placebo | .90 | 7 |
| Placebo | .95 | 8 |
| Alcohol | .98 | 9 |
| Alcohol | 1.11 | 10 |
| Alcohol | 1.27 | 11 |
| Alcohol | 1.32 | 12 |
| Alcohol | 1.44 | 13 |
| Alcohol | 1.45 | 14 |
| Alcohol | 1.46 | 15 |
| Placebo | 1.63 | 16 |
| Alcohol | 1.76 | 17 |
| Placebo | 1.97 | 18 |
| Alcohol | 2.56 | 19 |
| Alcohol | 3.07 | 20 |
|  |  |  |

## Drink driving reactions

$\mathrm{H}_{0}$ : There is no difference between the alcohol and placebo populations on reaction time
$\mathrm{H}_{\mathrm{a}}$ :
The alcohol population has a different reaction time distribution to the placebo population

Test Statistic $=$ Mann-Whitney U
The test statistic $U$ can be approximated to a $z$ score to get a $p$-value

## Mann-Whitney results

- Interpret the results

Test Statistics

|  | ReactionTime |
| :--- | ---: |
| Mann-Whitney U | 15.000 |
| $Z$ | -2.646 |
| p-value (2-tailed) | .008 |

Two tailed test


## Mann-Whitney results - solution

- Interpret the results
- $\mathrm{p}=0.008$
- Highly significant evidence to suggest a difference in the distributions of reaction times for those in the placebo and alcohol groups

Test Statistics

|  | ReactionTime |
| :--- | ---: |
| Mann-Whitney U | 15.000 |
| Z | -2.646 |
| p-value (2-tailed) | .008 |

Two tailed test


## ANOVA

## ANOVA

Compares the means of several groups
Which diet is best?
Dependent: Weight lost (Scale)
Independent: Diet 1, 2 or 3 (Nominal)
Null hypothesis: The mean weight lost on diets 1 , 2 and 3 is the same $\quad H_{0}: \mu_{1}=\mu_{2}=\mu_{3}$

Alternative hypothesis: The mean weight lost on diets 1,2 and 3 are not all the same

## Summary statistics

|  | Overall | Diet 1 | Diet 2 | Diet 3 |
| :--- | :--- | :--- | :--- | :--- |
| Mean | 3.85 | 3.3 | 3.03 | 5.15 |
| Standard deviation | 2.55 | 2.24 | 2.52 | 2.4 |
| Number in group | 78 | 24 | 27 | 27 |

- Which diet was best?
- Are the standard deviations similar?


## How Does ANOVA Work?

- $A N O V A=$ Analysis of variance
- We compare variation between groups relative to variation within groups
- Population variance estimated in two ways:
- One based on variation between groups we call the Mean Square due to Treatments/ MST/ MS between
- Other based on variation within groups we call the Mean Square due to Error/ MSE/ MS within


## Within group variation

Residual =difference between an individual and their group mean
$\mathrm{SS}_{\text {within }}=$ sum of squared residuals


## Between group variation

Differences between each group mean and the overall mean


## Sum of squares calculations

- $\mathrm{K}=$ number of groups

$$
\begin{aligned}
& S S_{\text {within }}=\sum_{j=1}^{k} \sum_{i=1}^{n}\left(x_{i j}-\overline{x_{j}}\right)^{2} \\
& =\sum_{i=1}^{24}\left(x_{i}-3.3\right)^{2}+\sum_{i=1}^{27}\left(x_{i}-3.03\right)^{2}+\sum_{i=1}^{27}\left(x_{i}-5.15\right)^{2}=430.179 \\
& \quad S_{\text {Between }}=\sum_{j=1}^{k} n_{j}\left(\overline{x_{j}}-\overline{x_{T}}\right)^{2} \\
& \quad=24(3.3-3.85)^{2}+27(3.03-3.85)^{2}+27(5.15-3.85)^{2}=71.094
\end{aligned}
$$

## ANOVA test statistic

Summary ANOVA


## Test Statistic (by hand)

- Filling in the boxes

|  | Sum of <br> squares | Degrees of <br> freedom | Mean <br> square | F-ratio (test <br> statistic) |
| :---: | :---: | :---: | :---: | :---: |
| SS $_{\text {between }}$ | 71.045 | 2 | 35.522 | 6.193 |
| SS $_{\text {within }}$ | 430.180 | 75 | 5.736 |  |
| SS $_{\text {total }}$ | 501.275 | 77 |  |  |

$F=\underline{\text { Mean between group sum of squared differences }}$ Mean within group sum of squared differences

If $F>1$, there is a bigger difference between groups than within groups

## P-value

- The p -value for ANOVA is calculated using the F distribution
- If you repeated the experiment numerous times, you would get a variety of test statistics



## One way ANOVA

$$
\text { Test Statistic }=\frac{\text { between group variation }}{\text { within group variation }}=\frac{\mathrm{MS}_{\text {Diet }}}{\mathrm{MS}_{\text {Error }}}=6.197
$$

Tests of Between-Subjects Effects
Dependent Variable: Weight lost on diet (kg)
$M S_{\text {between }}$
$M S_{\text {within }}$

| Source | Type III Sum of <br> Squares | df | Mean Square | F | Sig. |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Corrected Model | $71.094^{\mathrm{a}}$ | 2 | 35.547 | 6.197 | .003 |
| Intercept | 1137.494 | 1 | 1137.494 | 198.317 | .000 |
| Diet | 71.094 | 2 | 35.547 | 6.197 | .003 |
| Error | 430.179 | 75 | 5.736 |  |  |
| Total | 1654.350 | 78 |  |  |  |
| Corrected Total | 501.273 | 77 |  |  |  |

a. R Squared $=.142$ (Adjusted R Squared $=.119$ )

> There was a significant difference in weight lost between the diets $(p=0.003)$

Post hoc tests
If there is a significant ANOVA result, pairwise comparisons are made

They are t-tests with adjustments to keep the type 1 error to a minimum

- Tukey's and Scheffe's tests are the most commonly used post hoc tests.
- Hochberg's GT2 is better where the sample sizes for the groups are very different.


## Post hoc tests

- Which diets are significantly different?

Multiple Comparisons
Dependent Variable: Weight lost on diet (kg)

|  | (1) Diet | (J) Diet | Mean Difference (1J) | Std. Error | Sig. | 95\% Confidence Interval |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  | Lower Bound | Upper Bound |
| Tukey HSD | 1 | 2 | . 2741 | . 67188 | . 912 | -1.3325 | 1.8806 |
|  |  | 3 | -1.8481 ${ }^{\text {² }}$ | . 67188 | . 020 | -3.4547 | -. 2416 |
|  | 2 | 1 | -. 2741 | . 67188 | . 912 | -1.8806 | 1.3325 |
|  |  | 3 | $-2.1222^{*}$ | . 65182 | . 005 | -3.6808 | -. 5636 |
|  | 3 | 1 | $1.8481^{\prime \prime}$ | . 67188 | . 020 | . 2416 | 3.4547 |
|  |  | 2 | $2.1222^{*}$ | . 65182 | . 005 | . 5636 | 3.6808 |

- Write up the results and conclude with which diet is the best.


## Pairwise comparisons

- Results

| Test | $p$-value |
| :--- | :--- |
| Diet 1 vs Diet 2 |  |
| Diet 1 vs Diet 3 |  |
| Diet 2 vs Diet 3 |  |

- Report:


## Pairwise comparisons

- Results

| Test | P -value |
| :--- | :--- |
| Diet 1 vs Diet 2 | $\mathrm{P}=0.912$ |
| Diet 1 vs Diet 3 | $\mathrm{P}=0.02$ |
| Diet 2 vs Diet 3 | $\mathrm{P}=0.005$ |

There is no significant difference between Diets 1 and 2 but there is between diet 3 and diet $1(p=0.02)$ and diet 2 and diet $3(p=0.005)$.

The mean weight lost on Diets $1(3.3 \mathrm{~kg})$ and $2(3 \mathrm{~kg})$ are less than the mean weight lost on diet 3 ( 5.15 kg ).

## Assumptions for ANOVA

| Assumption | How to check | What to do if assumption <br> not met |
| :--- | :--- | :--- |
| Normality: The residuals <br> (difference between observed <br> and expected values) should be <br> normally distributed | Histograms/ <br> QQ plots/ <br> normality tests <br> of residuals | Do a Kruskall-Wallis test <br> which is non-parametric <br> (does not assume normality) |
| Homogeneity of variance (each <br> group should have a similar <br> standard deviation) | Levene's test | Welch test instead of <br> ANOVA and Games-Howell <br> for post hoc or Kruskall- <br> Wallis |

## Ex: Can equal variances be assumed?

, Null:

Levene's Test of Equality of Error Variancesa

Dependent Variable: WeightLOST

| F | df1 | df2 | Sig. |
| :---: | ---: | ---: | ---: |
| .659 | 2 | 75 | .520 |

Tests the null hypothesis that the error
variance of the dependent variable is equal across groups.
a. Design: Intercept + Diet
$p=$
Reject/ do not reject

## - Conclusion:

## Exercise: Can normality be assumed?



## Can normality be assumed?

Should you:<br>a) Use ANOVA

b) Use Kruskall-Wallis

## Ex: Can equal variances be assumed?

, Null:

Levene's Test of Equality of Error Variancesa

Dependent Variable: WeightLOST

| F | df1 | df2 | Sig. |
| :---: | ---: | ---: | ---: |
| .659 | 2 | 75 | .520 |

Tests the null hypothesis that the error
variance of the dependent variable is equal across groups.
a. Design: Intercept + Diet

- Conclusion: Equality of variances can be assumed


## Ex: Can normality be assumed?



## ANOVA

- Two-way ANOVA has 2 categorical independent between groups variables
e.g. Look at the effect of gender on weight lost as well as which diet they were on

|  | WeightLOST | Diet | gender |  |
| ---: | ---: | ---: | ---: | ---: |
| 16 | 1.1 | 1 | Male |  |
| 17 | 1.5 | 1 | Female |  |
| 18 | 1.7 | 2 | Female |  |
| 19 | 1.7 | 2 | Male |  |
| 20 | 1.9 | 1 | Female |  |
| 21 | 2.0 | 2 | Female |  |
| 22 | 2.0 | 2 | Female |  |
| 23 | 2.0 | 1 | Female |  |
| 24 | 2.0 | 1 | Female |  |

## Two-way ANOVA

- Dependent = Weight Lost
- Independents: Diet and Gender
- Tests 3 hypotheses:

1. Mean weight loss does not differ by diet
2. Mean weight loss does not differ by gender
3. There is no interaction between diet and gender

What's an interaction?

## Means plot

- Mean reaction times after consuming coffee, water or beer were taken and the results by drink or gender were compared.

| Mean Reaction times | male | female |
| :--- | ---: | ---: |
| alcohol | 30 | 20 |
| water | 15 | 9 |
| coffee | 10 | 6 |

## Means/ line/ interaction plot

- No interaction between gender and drink



## Means plot

- Interaction between gender and drink



## ANOVA

- Mixed between-within ANOVA includes some repeated measures and some between group variables
e.g. give some people margarine B instead of A and look at the change in cholesterol over time

| Repeated measures |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  |  |  | Margarine |
|  |  |  |  | A |
| 2 | 6.76 | 6.2 | 6.13 | A |
| 3 | 6.56 | 5.83 | 5.71 | A |
| 4 | 4.8 | 4.27 | 4.15 | B |
| 5 | 8.43 | 7.71 | 7.67 | A |
| 6 | 7.49 | 7.12 | 7.05 | B |
| 7 | 8.05 | 7.25 | 7.1 | A |
| 8 | 5.05 | 4.63 | 4.67 | A |

## Related resources

## Resources available on http://www.statstutor.ac.uk/

Paper scenario :Emmissions_scenario_Stcp-marshallowen-5b
Video scenario: :Mass_Customisation_Video_Stcp-marshallowen-1a
Top tips for stats support videos:
Careful_with_the_maths_Video_ Stcp-marshallowen-3a
Conjoint_Analysis_Video_Stcp-marshallowen-4a
Reference booklet: Tutors_quick_guide_to_statistics_7
SPSS training: Workbook: tutor_training_SPSS_workbook_6a, Solutions_for_tutor_training_SPSS_workbook_6b, Data_for_tutor_training_SPSS_workbook_6c

