

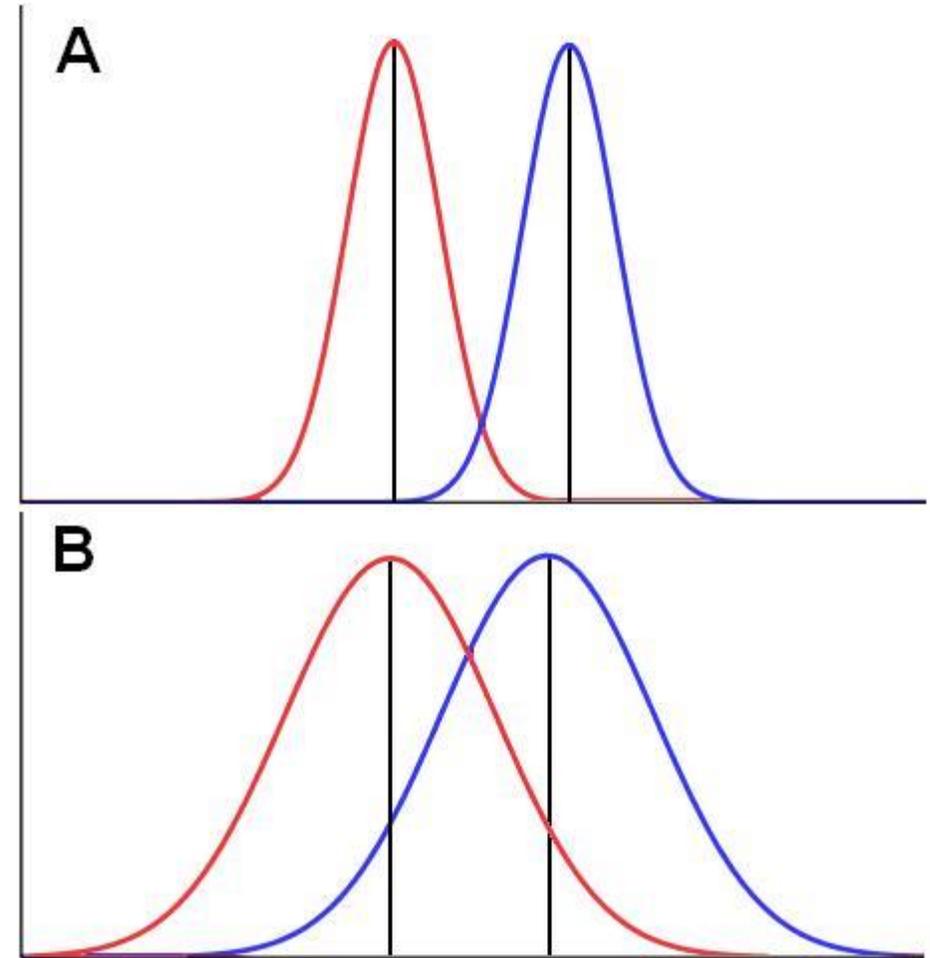
- <http://www.stefanokaburu.com/r-workshop.html>

Basic Statistics: SPSS

Dr. Colin Dubreuil

Comparing two groups

- A common activity in most fields of research is comparing 2 (or more) groups
- Many of our research questions are based around this concept
 - Do males and females of my study species differ in size?
 - Do males and females of my study species differ in terms of activity budget?
 - Do unaltered, natural bodies of water host a wider variety of species than anthropogenically modified bodies of water?
 - Do birds living near sources of anthropogenic noise produce songs that differ in structure as compared to birds living further away from anthropogenic noise?
 - Do animals produce more stress hormones before, or after a hurricane?



How do we 'know' if there is a difference between groups?

- We could compare the distributions visually
- We could calculate the means (or some other measure of central tendency) and see if they are the same

The issue

- The issue comes down to this: If I graph my data, or compare means, how do I know when a difference is different enough? Let's say I:
 - Count the number of water fowl species living in 7 'natural' bodies of water
 - Count the number of water fowl species living in 7 anthropogenically modified bodies of water
- Now, Calculate the means for each 'type' of habitat. I find that:
 - Natural bodies of water have (on average) 22.6 species of waterfowl (SD = 2.5)
 - Modified bodies of water have on average 20.1 species (SD = 6.1)
- **Is that different enough? Can I say my data suggest there is a real difference?**

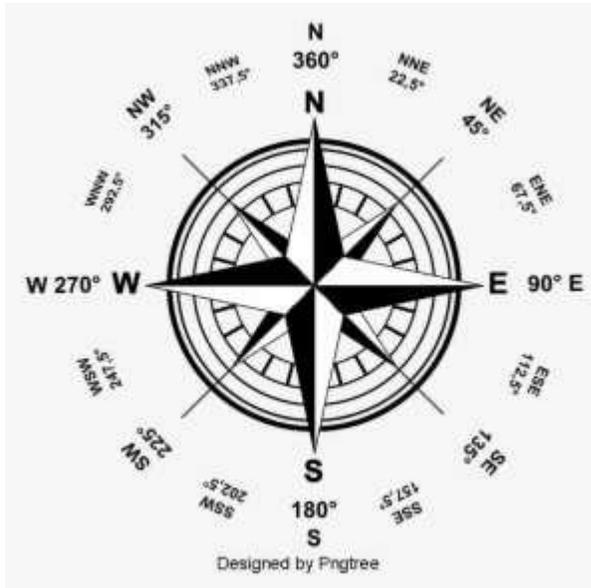
The issue

- The issue is that when we sample, there is a random aspect to it:
- Our sample statistics (mean of our sample, mode, etc...) are very likely to match the true population parameter perfectly
- You know that 10 flips of a coins should theoretically give you 5 heads, and 5 tails...
 - But it often doesn't because of the randomness of sampling

The issue

- **Take this silly question:**

- Are people who sit on the East side of graduate seminars generally taller than people who sit on the West side?



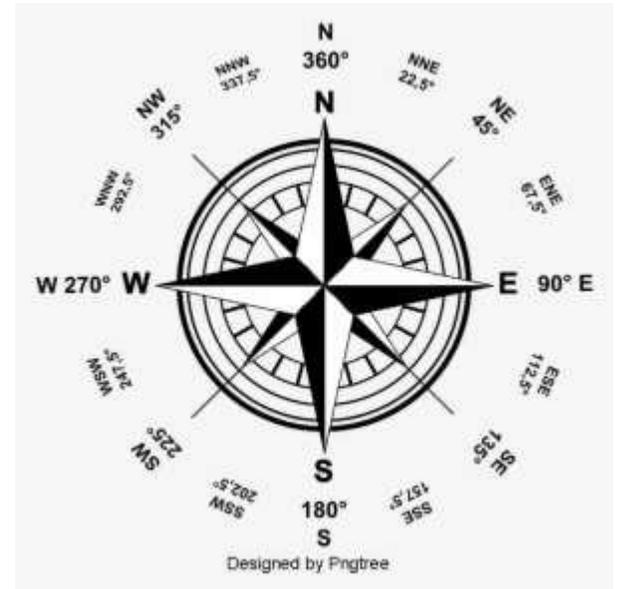
When is 'different' different enough?

- **Take this silly question:**

- Are people who sit on the 'east' side of graduate seminars generally taller than people who sit on the 'west' side?

- To answer this question, my best course of action would be to:

- Divide people in this room into 'east sitters', and 'west sitters'
- Take height measurements of people from each group
- Perhaps I could graph this data
- Take the mean (average) height of people in each group
- If I get different average values, have I proved anything?



The problem

- Likely not...
- In truth, I am likely to find a difference every time I do a comparison of 2 (or more) groups like this... Even if there isn't a 'real' difference
- I personally don't believe there is ANY relationship between height and whether you tend to sit on the 'east side' of a room vs the 'west side' ...
- But If I broke you all into those 2 groups, measured everyone in each group, and took the average, it's actually EXTREAMLY unlikely I would get the exact same average height for the 2 groups

A solution

- So, to determine if 2 groups are ‘different enough’ for us to consider them different, we tend to rely on **inferential statistical tests**
- Today, we will look at some **basic statistical tests** that will allow us to address the issue of: “Are these groups different enough to conclude there is a real world difference?”
 - The two sample T-Test
 - The non-parametric equivalent
 - The paired samples T-test
 - The non-parametric equivalent
 - Chi-squared test

The independent samples T-Test

- Inferential statistic used to determine if there is a significant difference between the means (the averages) of two groups.



An Example: Vervet body size dimorphism

- You are working in South Africa with a group of vervet monkeys
- Vervet monkeys live in multimale-multifemale groups
- You get the impression in your first few weeks in the field that adult male vervet monkeys are somewhat larger than females, but...
 - Based on visual inspection alone, you can't be certain
 - No one would EVER let you publish a paper that reveals there are sex differences in body size in vervets without some statistical evidence that this is true!



Vervet body size dimorphism

- You devise a plan to bring a scale into the field, bait it with honey, and get weights for individual monkeys
- After getting ethics approval, you move forward!



What a great field season!

- You now have data!
- Probably the first thing you should do is take a look at the data

- In SPSS, please open 'vervetweight.xlsx'



Visible: 0 of 0 Variables

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Open Data

Look in: Documents

- Custom Office Templates
- Fax
- Scanned Documents

File name:

Files of type: SPSS Statistics (*.sav, *.zsav)

Buttons: Open, Paste, Cancel, Help

Retrieve File From Repository...





Visible: 0 of 0 Variables

	var																		
1																			
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Open Data

Look in: Desktop

- Mammology
- Meeting with Stephano
- SPSS
- Undergrad supervision
- focalmaster.xlsx
- hm_interactions.xlsx
- monkey_attribute.xlsx
- vervetweight.xlsx

File name:

Files of type: Excel (*.xls, *.xlsx, *.xlsm)

Encoding:



	var	var	var	var	var
1					
2					
3					
4					
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22					
23					

Visible: 0 of 0 Variables

Read Excel File ✕

C:\Users\u26411\OneDrive - University of Wolverhampton\Desktop\wervetweight.xlsx

Worksheet: Sheet1 [A1:B22]

Range:

Read variable names from first row of data

Percentage of values that determine data type:

Ignore hidden rows and columns

Remove leading spaces from string values

Remove trailing spaces from string values

Preview

	Weight...	group
1	4.56	Male
2	5.47	Male
3	5.15	Male
4	5.23	Male
5	5.44	Male
6	5.13	Male
7	4.88	Male

i Final data type is based on all data and can be different from the preview, which is based on the first 200 data rows. The preview displays only the first 500 columns.

➔

OK
Paste
Reset
Cancel
Help



Visible: 2 of 2 Variables

	Weight_kg	group	var													
1	4.56	Male														
2	5.47	Male														
3	5.15	Male														
4	5.23	Male														
5	5.44	Male														
6	5.13	Male														
7	4.88	Male														
8	4.79	Male														
9	4.98	Male														
10	5.01	Male														
11	4.56	Female														
12	4.76	Female														
13	4.34	Female														
14	3.97	Female														
15	5.15	Female														
16	4.45	Female														
17	4.14	Female														
18	4.55	Female														
19	4.78	Female														
20	4.84	Female														
21	4.55	Female														
22																

Column: 'group'

- 10 Males
- 11 Females

Column: Weight_kg

- Weights of different individuals



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
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5	5.44	Male													
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8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Lets briefly explore our data!

Perhaps the first thing we can look into is the mean value of our groups...

Did males and females have a different weight on average?



Visible: 2 of 2 Variables

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16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Means

1

2

Weight_kg

group

Dependent List:

Layer 1 of 1

Previous Next

Independent List:

Options... Style... Bootstrap...

OK Paste Reset Cancel Help



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
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20	4.84	Female													
21	4.55	Female													
22															

Means

Dependent List:
Weight_kg

group

Layer 1 of 1

Independent List:

Options...
Style...
Bootstrap...

Previous Next

OK Paste Reset Cancel Help



Visible: 2 of 2 Variables

	Weight_kg	group	var												
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20	4.84	Female													
21	4.55	Female													
22															

Means

Dependent List:
Weight_kg

Layer 1 of 1
Previous Next

Independent List:

OK Paste Reset Cancel Help

Options...
Style...
Bootstrap...



Visible: 2 of 2 Variables

	Weight_kg	group	var												
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17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Means

1

group

Dependent List:
Weight_kg

2

Independent List:

Options...
Style...
Bootstrap...

OK Paste Reset Cancel Help



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
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19	4.78	Female													
20	4.84	Female													
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22															

Means

Dependent List:
Weight_kg

Layer 1 of 1
Previous Next

Independent List:
group

Options...
Style...
Bootstrap...

OK Paste Reset Cancel Help





Visible: 2 of 2 Variables

	Weight_kg	group	var												
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17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Means

Dependent List:
Weight_kg

Layer 1 of 1
group

Options...
Style...
Bootstrap...

Previous Next

OK Paste Reset Cancel Help



	Weight_kg	group	var	var	var	var
1	4.56	Male				
2	5.47	Male				
3	5.15	Male				
4	5.23	Male				
5	5.44	Male				
6	5.13	Male				
7	4.88	Male				
8	4.79	Male				
9	4.98	Male				
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14	3.97	Female				
15	5.15	Female				
16	4.45	Female				
17	4.14	Female				
18	4.55	Female				
19	4.78	Female				
20	4.84	Female				
21	4.55	Female				
22						

Statistics:

- Median
- Grouped Median
- Std. Error of Mean
- Sum
- Minimum
- Maximum
- Range
- First
- Last
- Variance
- Kurtosis
- Std. Error of Kurtosis
- Skewness
- Std. Error of Skewness

Cell Statistics:

- Number of Cases
- Mean
- Standard Deviation

Statistics for First Layer

- Anova table and eta
- Test for linearity

Continue Cancel Help



Select the measurements you are interested in from the 'Statistics' box

Move them over to the 'Cell statistics' box with the arrow button

When you are done, click continue



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
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17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Means

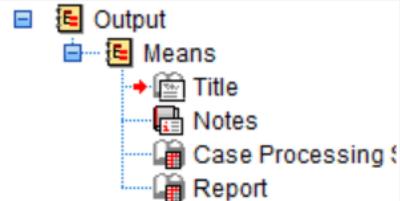
Dependent List:
Weight_kg

Layer 1 of 1
group

Options...
Style...
Bootstrap...

Previous Next

OK Paste Reset Cancel Help



Means

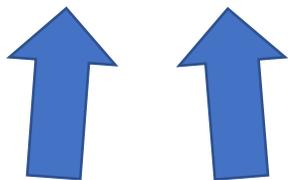
Case Processing Summary

	Included		Cases Excluded		Total	
	N	Percent	N	Percent	N	Percent
Weight_kg * group	21	100.0%	0	0.0%	21	100.0%

Report

Weight_kg group	N	Mean	Std. Deviation
Female	11	4.5536	.33161
Male	10	5.0640	.28254
Total	21	4.7967	.39890

Double-click to activate

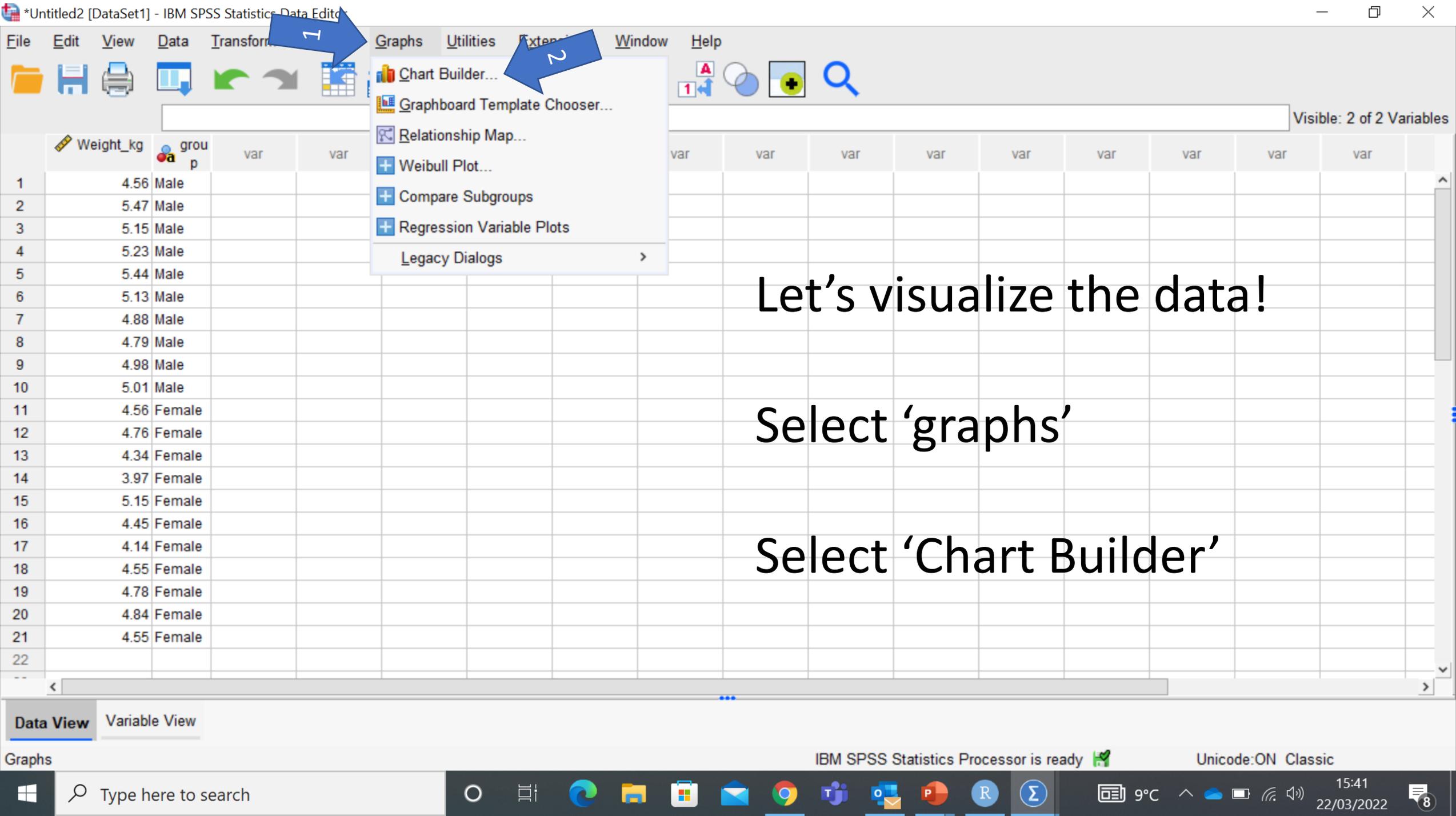


Notice here, we have counts by group

- We have 11 females in our sample
- We have 10 males in our sample
- A total of 21 individuals in our sample!

We also have mean weight by group:

- Females have an average weight of 4.55kg
- Males have an average weight of 5.06kg
- The average weight of all monkeys is 4.80kg
- Based on this, it looks like males *might* be heavier than females... But there is more we can do to investigate this properly!



Let's visualize the data!

Select 'graphs'

Select 'Chart Builder'



	Weight_kg	group	var								
1	4.56	Male									
2	5.47	Male									
3	5.15	Male									
4	5.23	Male									
5	5.44	Male									
6	5.13	Male									
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13	4.34	Female									
14	3.97	Female									
15	5.15	Female									
16	4.45	Female									
17	4.14	Female									
18	4.55	Female									
19	4.78	Female									
20	4.84	Female									
21	4.55	Female									
22											

Chart Builder

Before you use this dialog, measurement level should be set properly for each variable in your chart. In addition, if your chart contains categorical variables, value labels should be defined for each category.

Press OK to define your chart.

Press Define Variable Properties to set measurement level or define value labels for chart variables.

Don't show this dialog again

OK **Define Variable Properties...**



This warning dialogue may come up

It is telling you to be sure your data are set to the correct 'measurement level'

Remember that sex is a categorical variable. In SPSS this is coded as 'Nominal'

Weight is continuous
In SPSS, this is coded as 'Scale'

Variables: *Chart preview uses example data*

Weight_kg
group

No categories (scale variable)

Drag a Gallery chart here to use it as your starting point

OR

Click on the Basic Elements tab to build a chart element by element

Gallery Basic Elements Groups/Point ID Titles/Footnotes

Choose from:

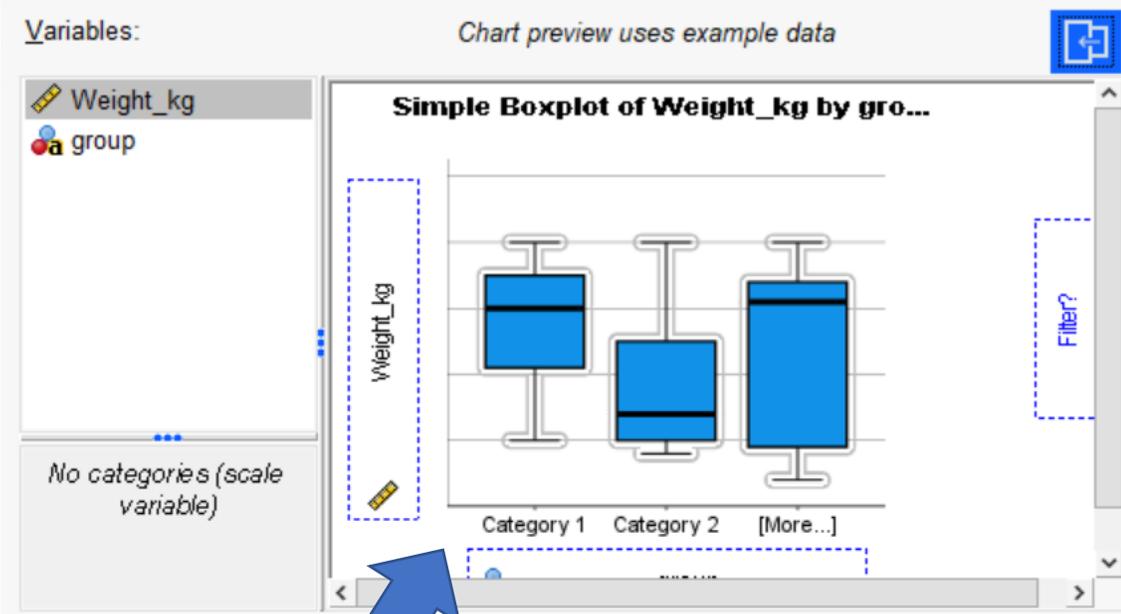
- Favorites
- Bar
- Line
- Area
- Pie/Polar
- Scatter/Dot
- Histogram
- High-Low
- Boxplot
- Dual Axes

OK Paste Reset Cancel Help

- This should bring up a window that looks something like this...
- We want to compare the value of a continuous variable across different categories
- A box plot is our best friend for this!



	Weight_kg	group
1	4.56	Male
2	5.47	Male
3	5.15	Male
4	5.23	Male
5	5.44	Male
6	5.13	Male
7	4.88	Male
8	4.79	Male
9	4.98	Male
10	5.01	Male
11	4.56	Fema
12	4.76	Fema
13	4.34	Fema
14	3.97	Fema
15	5.15	Fema
16	4.45	Fema
17	4.14	Fema
18	4.55	Fema
19	4.78	Fema
20	4.84	Fema
21	4.55	Fema
22		



Gallery Basic Elements Groups/Point ID Titles/Footnotes

Choose from:

- Favorites
- Bar
- Line
- Area
- Pie/Polar
- Scatter/Dot
- Histogram
- High-Low
- Boxplot**
- Dual Axes

OK Paste Reset Cancel Help

Element Properties Chart Appearance Options

Edit Properties of:

- Box1
- X-Axis1 (Box1)
- Y-Axis1 (Box1)

Statistics

Variable: Weight_kg

Statistic: Boxplot

Set Parameters...

Display error bars

Error Bars Represent

- Confidence intervals
 - Level (%): 95
- Standard error
 - Multiplier: 2
- Standard deviation

Step 1: In the lower box, select 'Box Plot'

Step 2, drag the basic boxplot from the images that pop up, and drag it above into the 'Drag gallery chat here...' box



	Weight_kg	g
1	4.56	Male
2	5.47	Male
3	5.15	Male
4	5.23	Male
5	5.44	Male
6	5.13	Male
7	4.88	Male
8	4.79	Male
9	4.98	Male
10	5.01	Male
11	4.56	Fema
12	4.76	Fema
13	4.34	Fema
14	3.97	Fema
15	5.15	Fema
16	4.45	Fema
17	4.14	Fema
18	4.55	Fema
19	4.78	Fema
20	4.84	Fema
21	4.55	Fema
22		

Variables: *Chart preview uses example data*

Weight_kg

group

Simple Boxplot of Weight_kg by gro...

Category 1 Category 2 [More...]

group

Gallery Basic Elements Groups/Point ID Titles/Footnotes

Choose from:

- Favorites
- Bar
- Line
- Area
- Pie/Polar
- Scatter/Dot
- Histogram
- High-Low
- Boxplot
- Dual Axes

OK Paste Reset Cancel Help

Element Properties Chart Appearance Options

Edit Properties of:

Box1

X-Axis1 (Box1)

Y-Axis1 (Box1)

Statistics

Variable: Weight_kg

Statistic:

Boxplot

Set Parameters...

Display error bars

Error Bars Represent

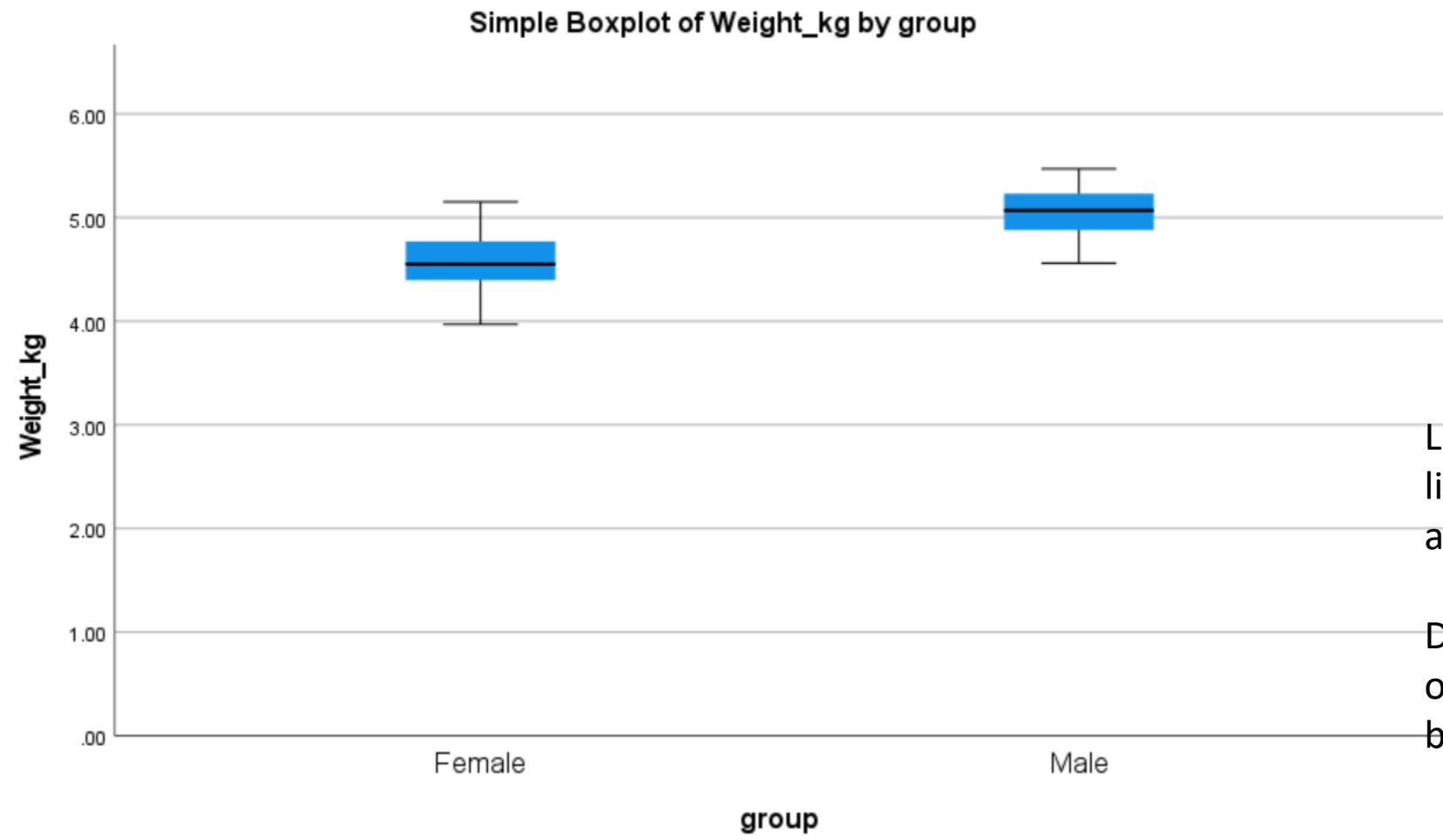
Confidence intervals

- In a standard boxplot:
- We drag our grouping variable (the categorical one) to the x-axis
- We drag our continuous variable to the y-axis
- Once this is done, you can click 'ok'



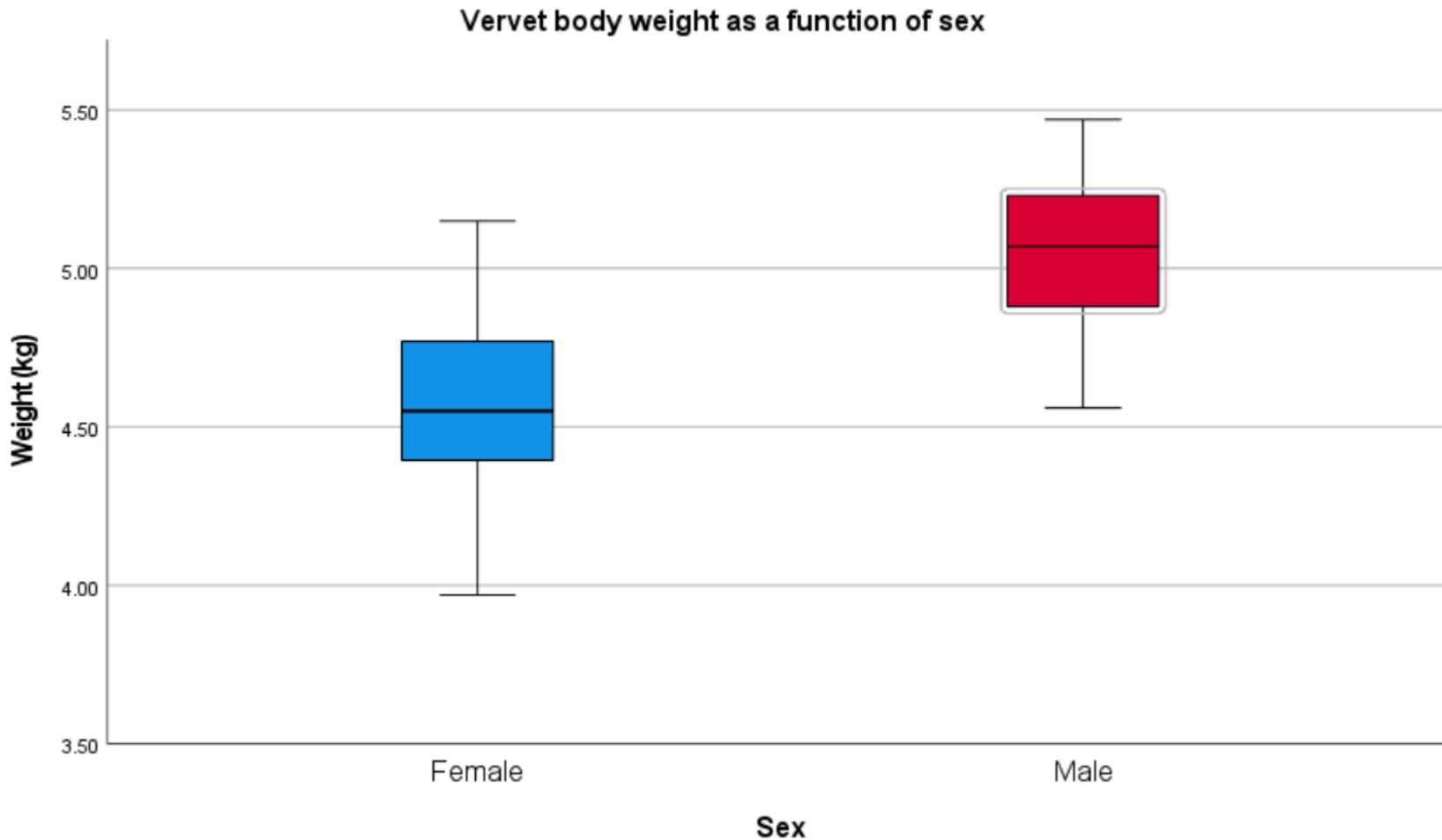
- Output
 - Means
 - Title
 - Notes
 - Case Processing S...
 - Report
 - GGraph
 - Title
 - Notes
 - Graph

→ GGraph



Let's make this a little nicer to look at!

Double click it to open a dialogue box



- That's better
- Hey, this figure seems to line up nicely with our inference from the means:
 - Males seem to be heavier than females!

- Ok. So we have different mean values for the weight of males vs female vervet monkeys
- Our figure seems to back up the idea that males are heavier than females
- But is there a statistically significant difference between the males and females?
 - For this, we will explore the **independent samples T Test**

Independent samples T Test

- A test used to see if there is a significant difference between the **means of two groups**
 - You can start thinking about T Tests when you have 2 groups
 - The variable that represents the 'groups' is by definition a categorical variable
 - You have a second variable that you think might vary by group, and that variable is continuous
- **The null hypothesis:** the 2 population (group) means are equal
- **Alternative hypothesis:** the 2 population (group) means are different

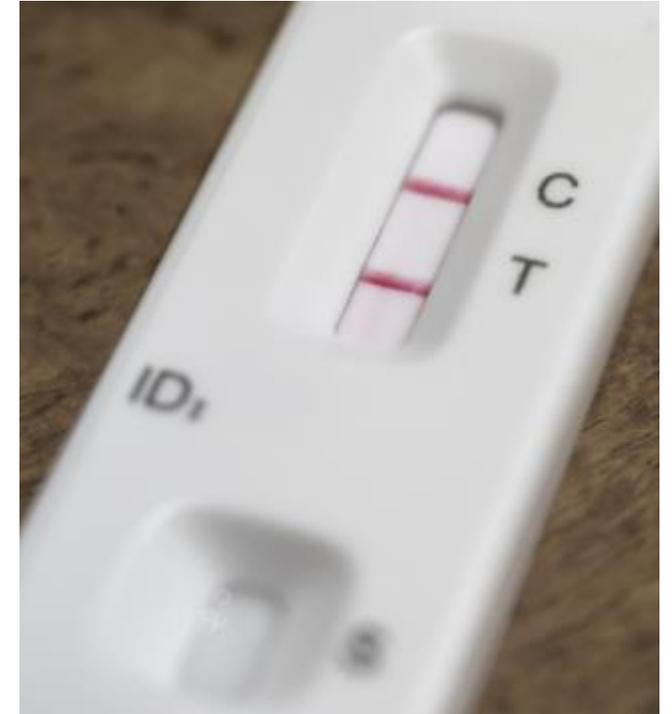
IMPORTANT: Assumptions

- Any statistical test is a procedure that can tell us something about our data, provided certain **assumptions** are met...
- Results from a test cannot be interpreted properly if basic assumptions are not met!
- Let's take/use a COVID rapid antigen test as a metaphor

IMPORTANT: Assumptions

- If you showed me a COVID test you brought from home, and it is displaying a positive result, I would believe you had COVID, as long as some ASSUMPTIONS were met
 - Did YOU take the test, or did you swab our brother?
 - Did you take the test recently, or was that test taken 3 weeks ago?
 - Did you follow the instructions on the box?
- Assuming these ASSUMPTIONS are met, I would infer from a positive COVID test that you have COVID...
- Otherwise, that test doesn't mean much!

- The same logic holds true for any statistical test!



Assumptions of an Two Samples T Test

Independence of observations

Normality

No significant Outliers

Homogeneity of Variance

Assessing assumptions

- Independence
 - Here, you are really assessing the **research design**
 - Was the data sampled randomly?
 - Can you consider the data points independent of one another?
- Possible independence issues to consider:
 - If you took all your “female” measurements from a group of related females
 - Did you accidentally collect male weights before a drought, and then females during the drought?
- This assumption needs to be considered thoroughly!
 - If the assumption is not met, **there are other, more appropriate statistics you can use!**

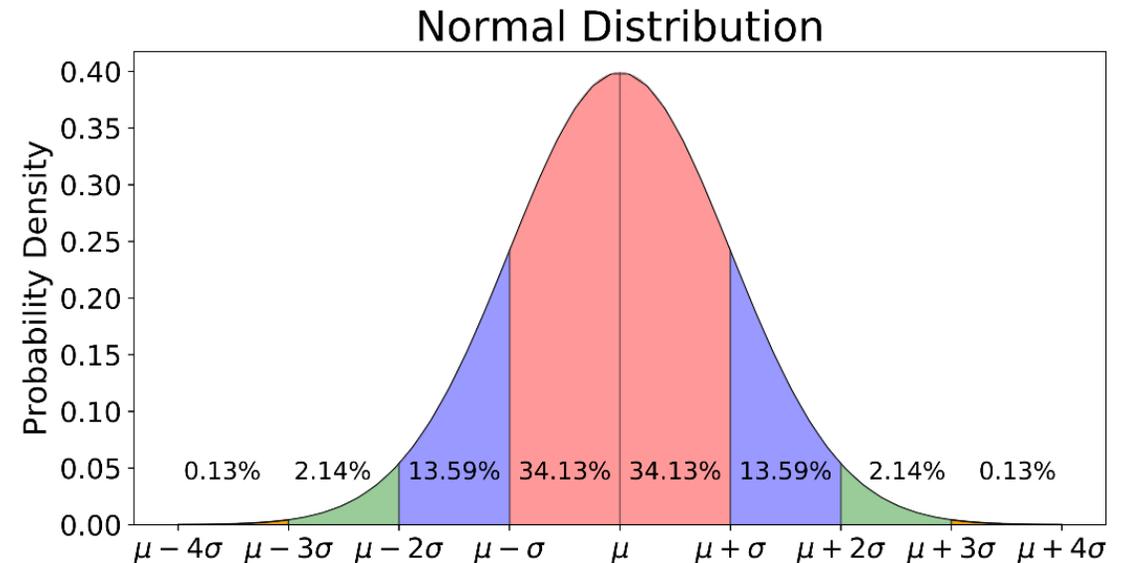
Normality + Outliers

- The distribution of the data **in both your groups** should be 'normal'
- There are some specific mathematical descriptions of what makes a **normal distribution**, but for our purposes today:



- The Normal distribution (AKA Gaussian distribution), is a distribution where data is spread symmetrical around the mean

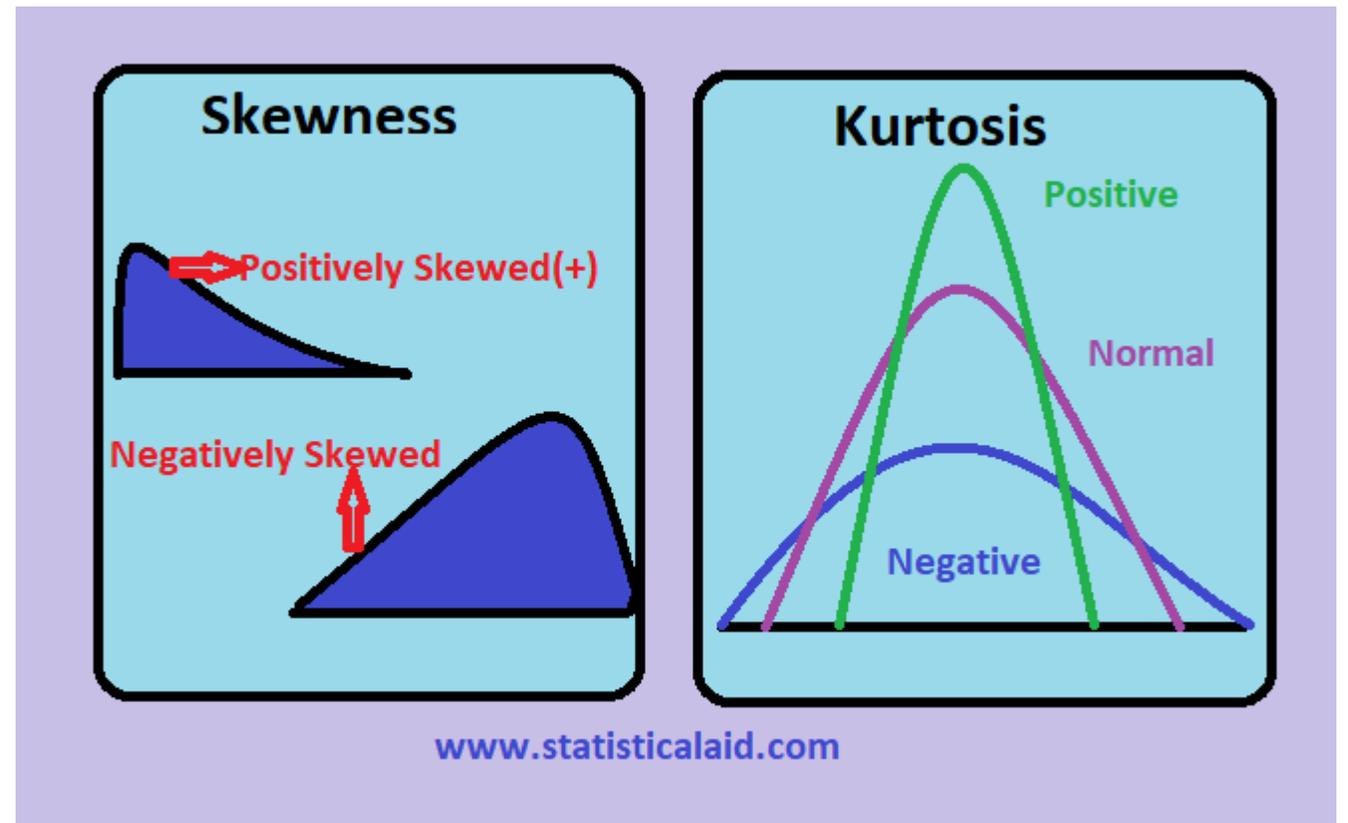
- The bulk of the data is closer to the mean, and tapers off as you move away from the mean.
- The standard 'bell curve'.



- ~68% of the data should be with 1 SD from the mean
- ~95% of the data should be with 2 SD of the mean
- ~99.7% of the data should be within 3 SD of the mean

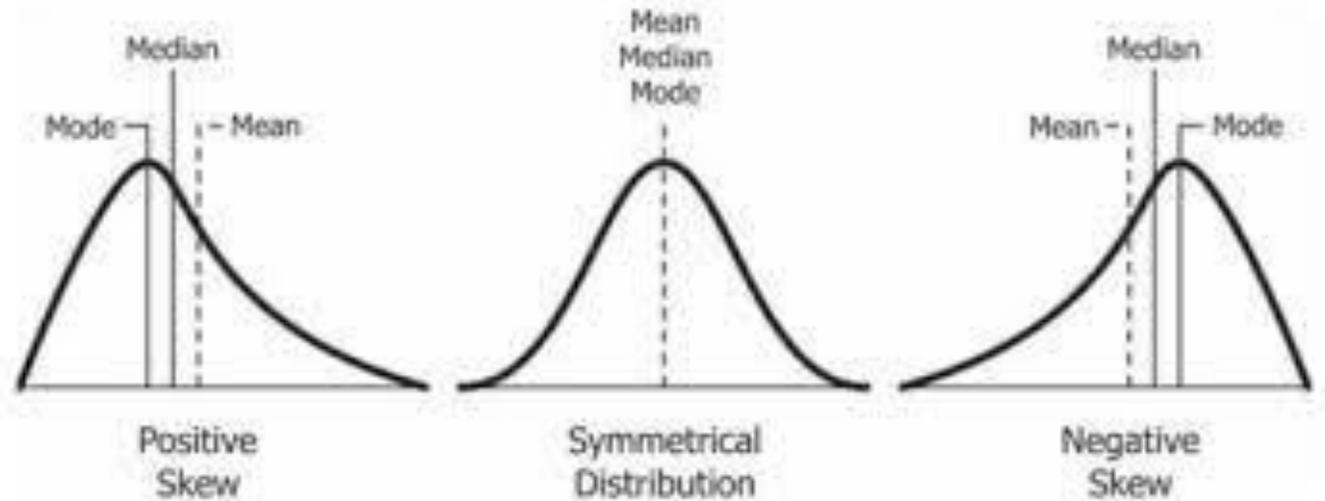
Normality + Outliers

- There are some very specific criteria that must be met for a 'bell curve' to be considered normally distributed
- One way of checking for normality is to look at **skewness** and **kurtosis**



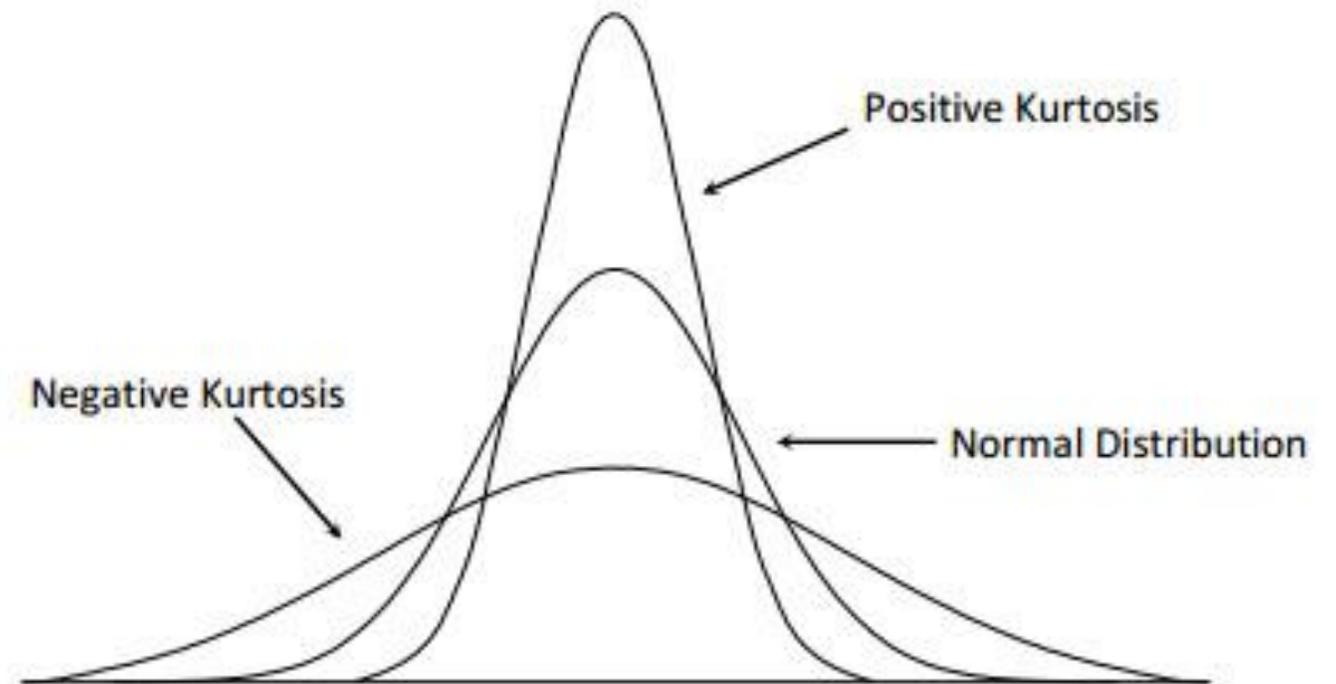
Skewness

- **Skewness:** a measure of symmetry
 - **Positively skewed** distributions have long tails to the right (positive side) of the distribution
 - **Negatively skewed** distributions have long tails to the left (negative side) of the distribution
- Normal distributions should have low (close to zero) levels of skewness



Kurtosis

- **Kurtosis:** How 'fat' are the tails of your distribution
- Normal distributions have kurtosis values close to zero
 - If your data have **positive kurtosis values**, sharper peak and heavier tails compared to a normal distribution
 - **Negative kurtosis values** suggest flatter peak and thinner tails compared to a normal distribution



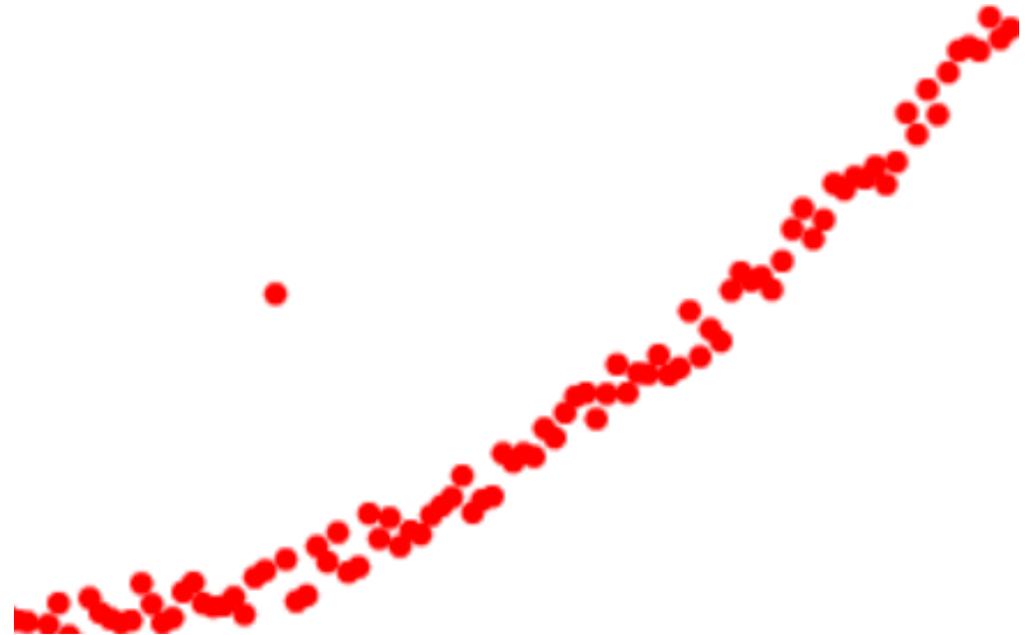
Assessing Normality: Skewness and Kurtosis

- Without getting too math-y: As a rule of thumb, you are looking for values of kurtosis and skewness that are **between -2 and +2 (zero would be optimum)**
- Normality can also be assessed using a statistical test: **Shapiro-Wilk**
- We also assess normality using **qqplots**.

- But wait... what about outliers?!

Normality + Outliers

- Simply put, an outlier is an extremely high or extremely low data point relative to the rest of your data set
- Outliers are extreme values that stand out greatly from the overall pattern of values in a dataset or graph.



Assessing Normality + Outliers in SPSS = Easy!

- Great news! SPSS can help us look at Both Normality AND Outliers in one output!



Visible: 2 of 2 Variables

	Weight_kg	group	var													
1	4.56	Male														
2	5.47	Male														
3	5.15	Male														
4	5.23	Male														
5	5.44	Male														
6	5.13	Male														
7	4.88	Male														
8	4.79	Male														
9	4.98	Male														
10	5.01	Male														
11	4.56	Female														
12	4.76	Female														
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17	4.14	Female														
18	4.55	Female														
19	4.78	Female														
20	4.84	Female														
21	4.55	Female														
22																

Data View Variable View



Visible: 2 of 2 Variables

	Weight_kg	group	var											
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18	4.55	Female												
19	4.78	Female												
20	4.84	Female												
21	4.55	Female												
22														

Explore

Weight_kg → Dependent List

group → Factor List

Statistics... Plots... Options... Bootstrap...

Label Cases by:

Display: Both Statistics Plots

OK Paste Reset Cancel Help

Move 'Weight_kg' to the dependent list

Move 'group' to the Factor List



Visible: 2 of 2 Variables

	Weight_kg	group	var											
1	4.56	Male												
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3	5.15	Male												
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17	4.14	Female												
18	4.55	Female												
19	4.78	Female												
20	4.84	Female												
21	4.55	Female												
22														

Explore

Dependent List:
Weight_kg

Factor List:
group

Label Cases by:

Display
 Both Statistics Plots

OK Paste Reset Cancel Help

Statistics...
Plots...
Options...
Bootstrap...



Click 'Statistics'



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
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17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Explore

Dependent List:

Explore: Statistics

- Descriptives
Confidence Interval for Mean: 95 %
- M-estimators
- Outliers
- Percentiles

Display: Both Statistics Plots

Buttons: Continue, Cancel, Help

Check 'Outliers', and then click 'Continue'



Visible: 2 of 2 Variables

	Weight_kg	group	var											
1	4.56	Male												
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17	4.14	Female												
18	4.55	Female												
19	4.78	Female												
20	4.84	Female												
21	4.55	Female												
22														

Explore

Dependent List:
Weight_kg

Factor List:
group

Label Cases by:

Display
 Both Statistics Plots

OK Paste Reset Cancel Help

Statistics...
Plots...
Options...
Bootstrap...



Click 'Plots'



Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
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17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Explore: Plots

Boxplots

- Factor levels together
- Dependents together
- None

Descriptive

- Stem-and-leaf
- Histogram

Normality plots with tests

Spread vs Level with Levene Test

- None
- Power estimation
- Transformed Power: Natural log
- Untransformed

Continue Cancel Help

We want to see:

- Histogram
- Normality plots with tests
- You can uncheck "stem-and-leaf", although it wont hurt if you leave it checked



Visible: 2 of 2 Variables

	Weight_kg	group	var												
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17	4.14	Female													
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19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Explore

Dependent List:
Weight_kg

Factor List:
group

Label Cases by:

Display
 Both Statistics Plots

OK Paste Reset Cancel Help

Statistics... Plots... Options... Bootstrap...





Output

- Explore
 - Title
 - Notes
 - Active Dataset
 - group
 - Title
 - Case Process
 - Descriptives
 - Tests of Norm
 - Weight_kg
 - Title
 - Histogram
 - Title
 - group
 - group
 - Normal Q-Q
 - Title
 - group
 - group
 - Detrended
 - Title
 - group
 - group
 - Boxplot

group		Statistic	Std. Error		
Weight_kg	Female	Mean	4.5536	.09998	
		95% Confidence Interval for Mean	Lower Bound	4.3309	
		Upper Bound	4.7764		
	5% Trimmed Mean	4.5529			
	Median	4.5500			
	Variance	.110			
	Std. Deviation	.33161			
	Minimum	3.97			
	Maximum	5.15			
	Range	1.18			
	Interquartile Range	.44			
	Skewness	-.082	.661		
	Kurtosis	.157	1.279		
Male	Mean	Mean	5.0640	.0935	
		95% Confidence Interval for Mean	Lower Bound	4.8619	
		Upper Bound	5.2661		
	5% Trimmed Mean	5.0694			
	Median	5.0700			
	Variance	.080			
	Std. Deviation	.28254			
	Minimum	4.56			
	Maximum	5.47			
	Range	.91			
	Interquartile Range	.43			
	Skewness	-.185	.687		
	Kurtosis	-.251	1.334		

For this test, we need to check for normality and outliers in BOTH our groups:

ie., in Females AND in males **separately**

I see here that for both sexes, skewness and kurtosis are between 2 and -2 for both males and females!

Great!



Output

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 - group
 - Normal Q
 - Title
 - group
 - group
 - Detrende
 - Title
 - group
 - group
 - Boxplot

Variance	.080	
Std. Deviation	.28254	
Minimum	4.56	
Maximum	5.47	
Range	.91	
Interquartile Range	.43	
Skewness	-.185	.687
Kurtosis	-.251	1.334

Tests of Normality

group	Kolmogorov-Smirnov ^a			Shapiro-Wilk		
	Statistic	df	Sig.	Statistic	df	Sig.
Weight_kg Female	.132	11	.200*	.979	11	.959
Weight_kg Male	.108	10	.200*	.974	10	.927

*. This is a lower bound of the true significance.
 a. Lilliefors Significance Correction

Weight_kg

Histograms

Histogram
for group= Female



The **Shapiro-Wilk test** is a test for normality

When this test is significant (ie., the p value is BELOW 0.05), it means our distribution is statistically DIFFERENT than normal

We want a bigger value here, and we have it for both males AND females!!! Good news!

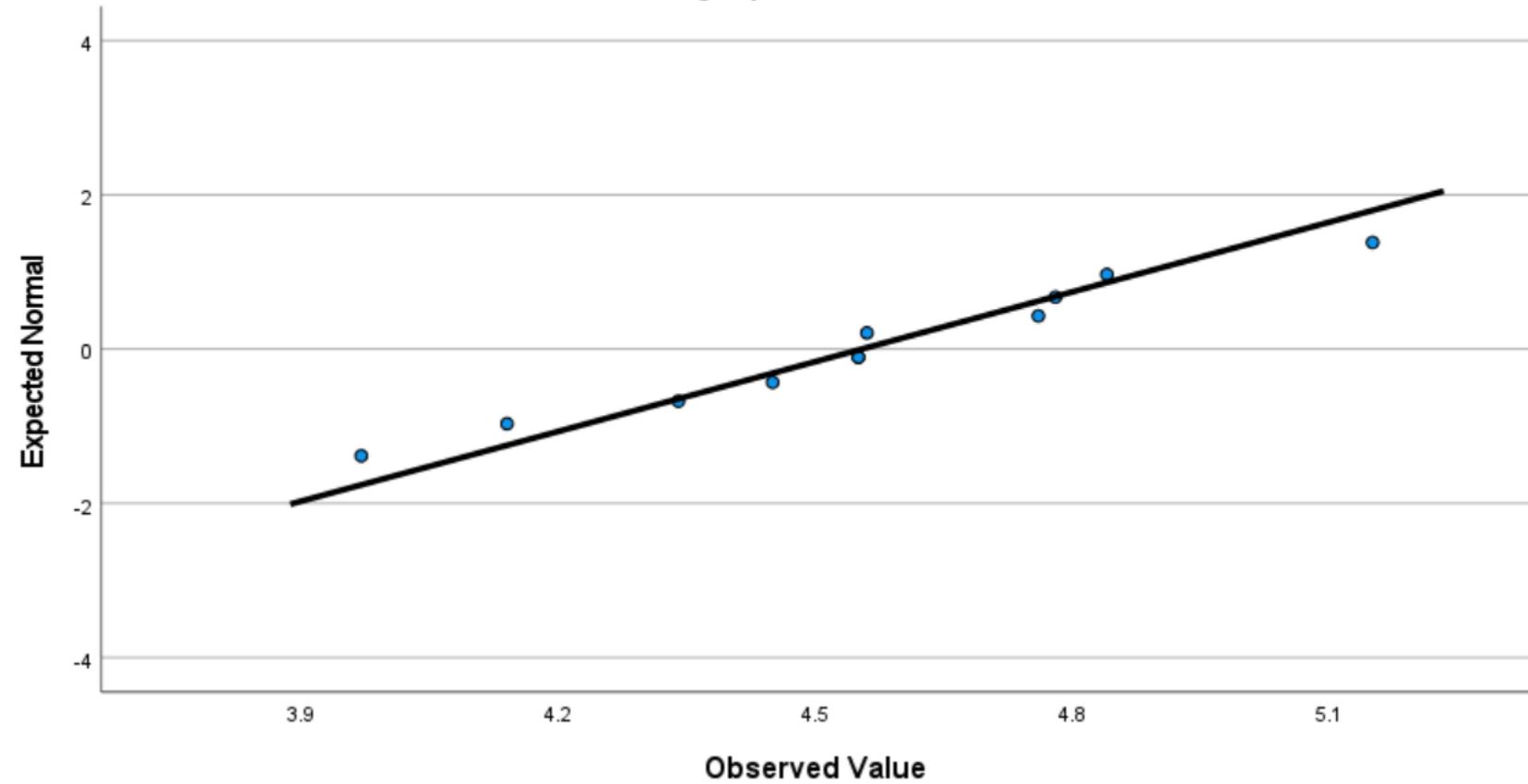
Small sample size: Makes it easier to get a non-significant result. Our sample is small, so we want to look at more than this



- Output
 - Explore
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 - Title
 - Case Process
 - Descriptives
 - Tests of Norm
 - Weight_kg
 - Title
 - Histogram
 - Title
 - group
 - group
 - Normal Q
 - Title
 - group
 - group
 - Detrende
 - Title
 - group
 - group
 - Boxplot

Normal Q-Q Plots

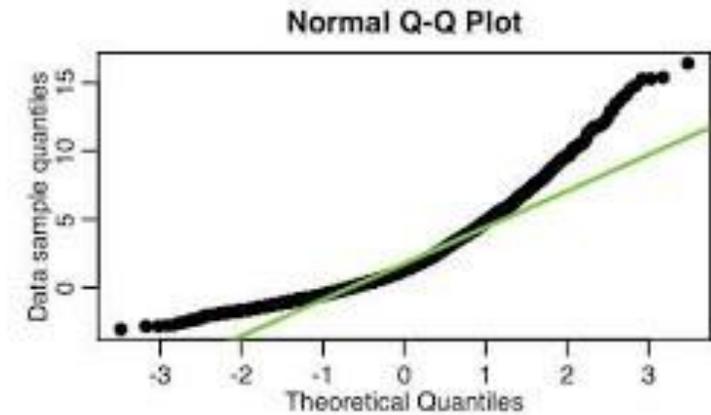
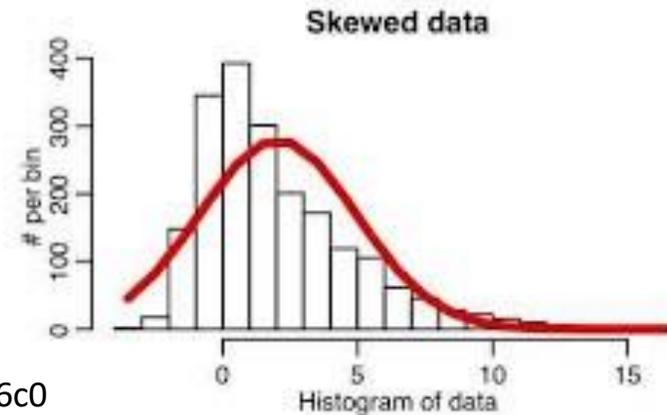
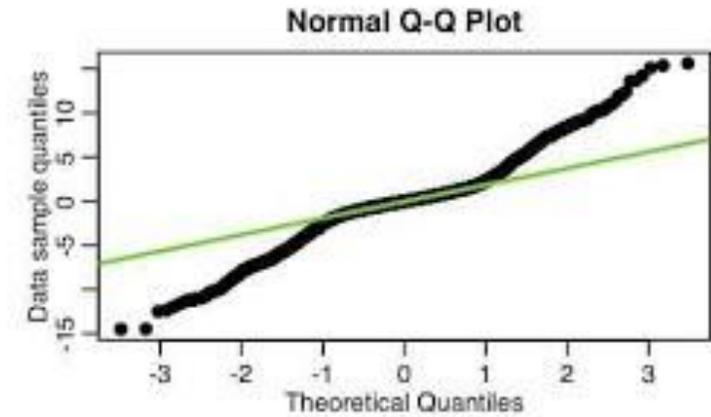
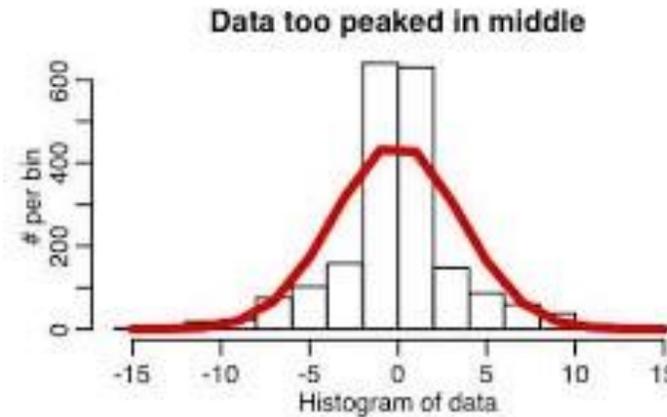
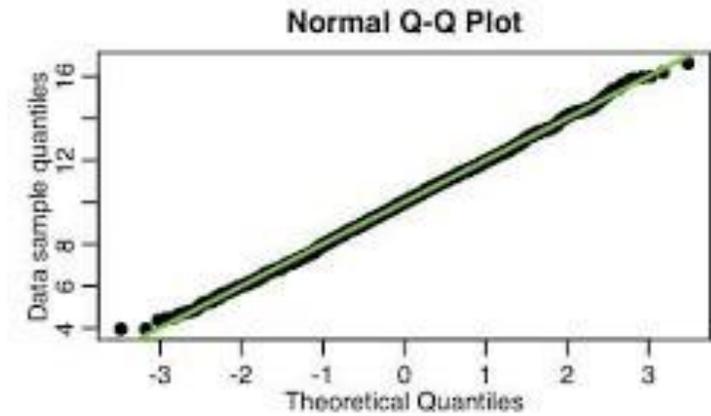
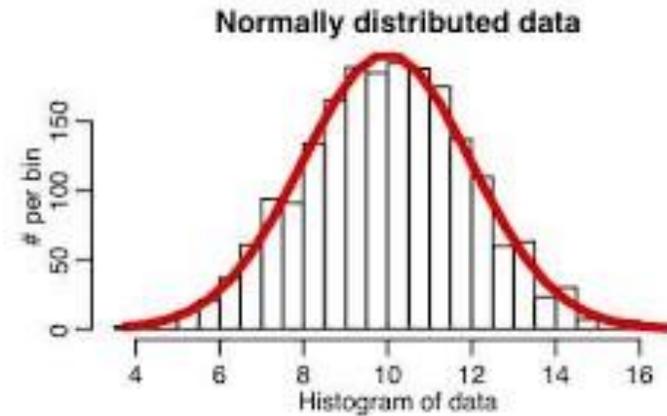
Normal Q-Q Plot of Weight_kg for group= Female



Double-click to activate

qqplots

- If data are normally distributed, you should see the points generally follow the straight line on the plot
- Having one or two points being off the line is **fine**
- **A trend, or pattern** in the points away from the line indicates a lack of normality

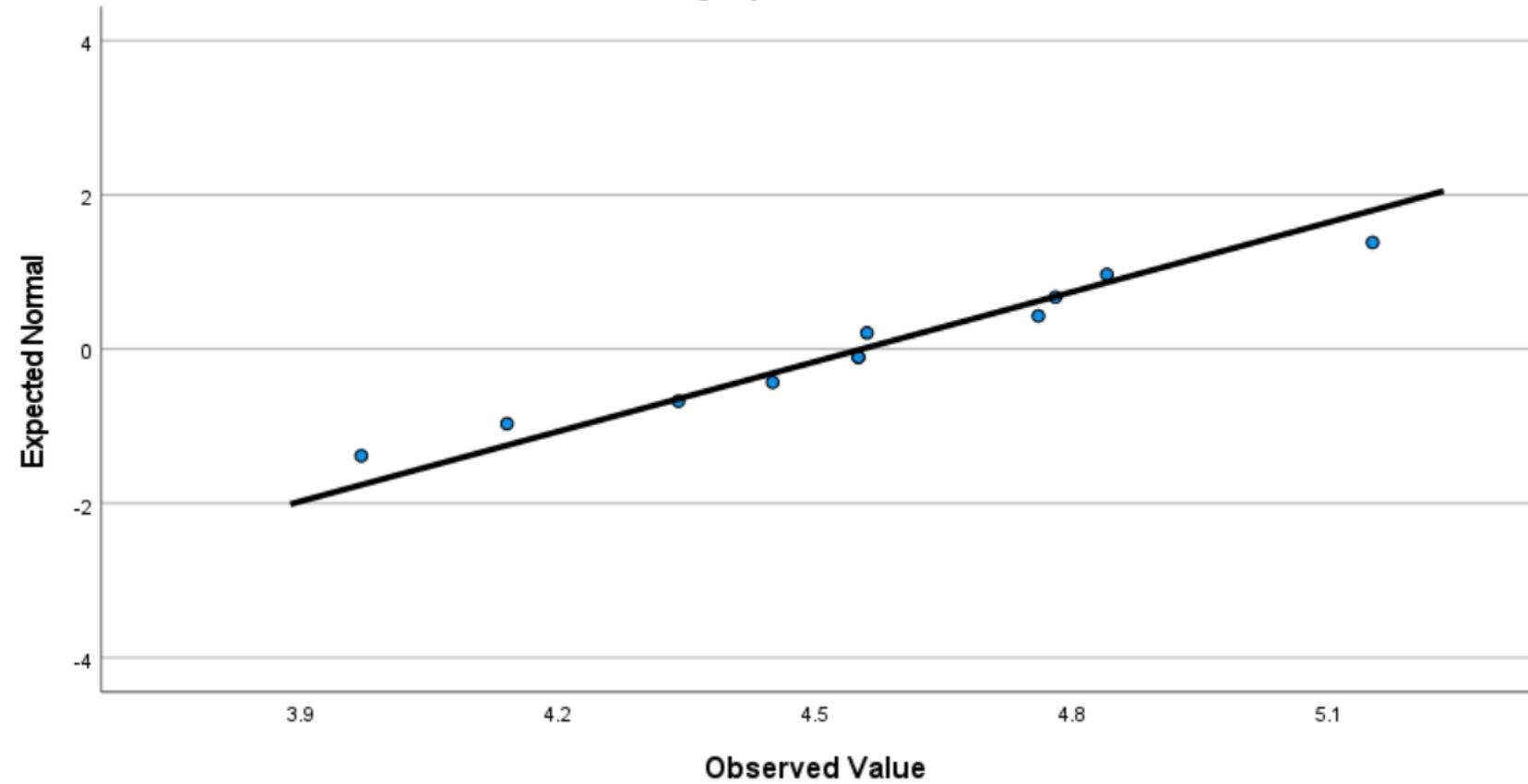




- Output
 - Explore
 - Title
 - Notes
 - Active Dataset
 - group
 - Title
 - Case Process
 - Descriptives
 - Tests of Norm
 - Weight_kg
 - Title
 - Histogram
 - Title
 - group
 - group
 - Normal Q
 - Title
 - group
 - group
 - Detrende
 - Title
 - group
 - group
 - Boxplot

Normal Q-Q Plots

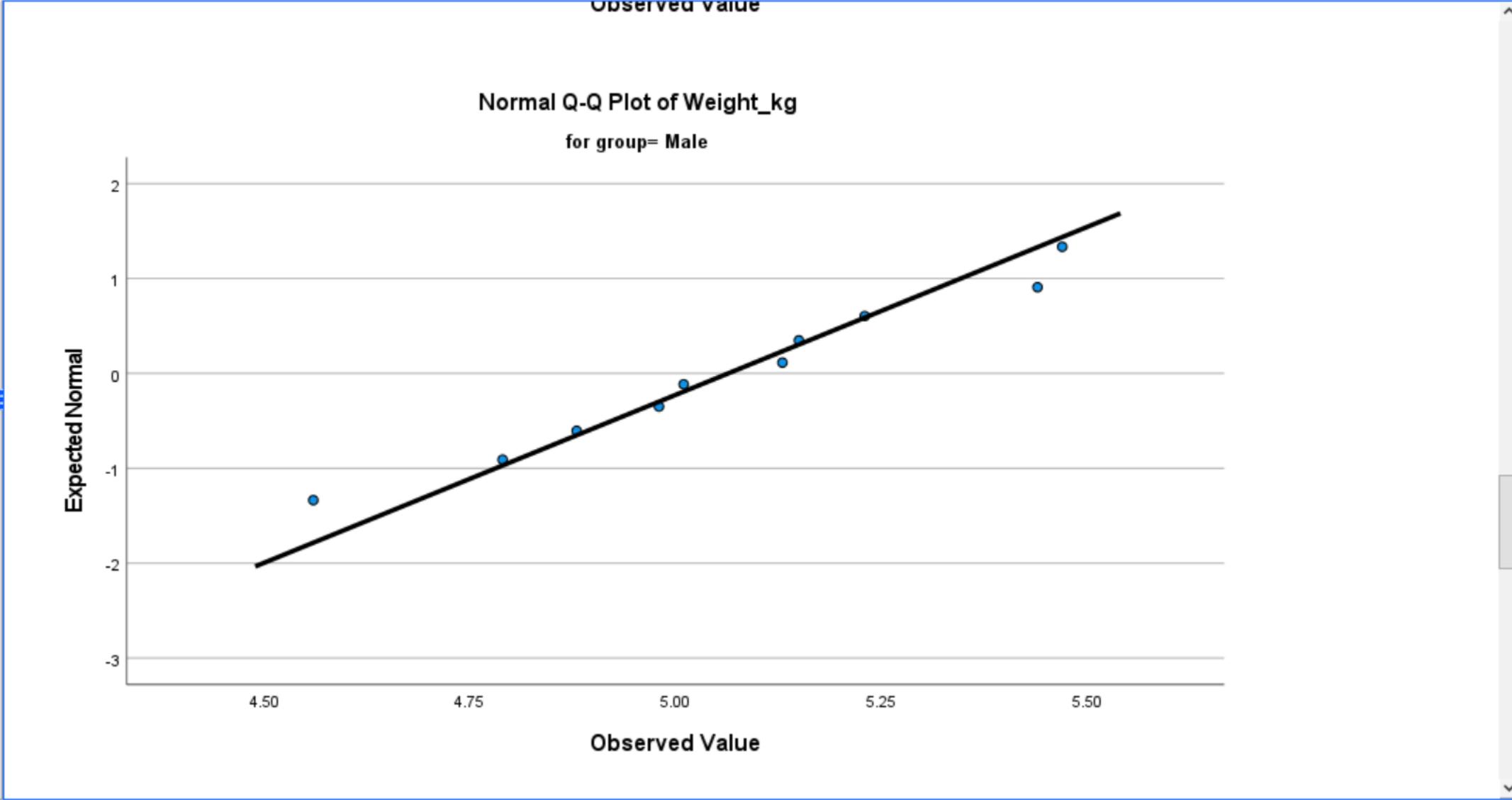
Normal Q-Q Plot of Weight_kg for group= Female



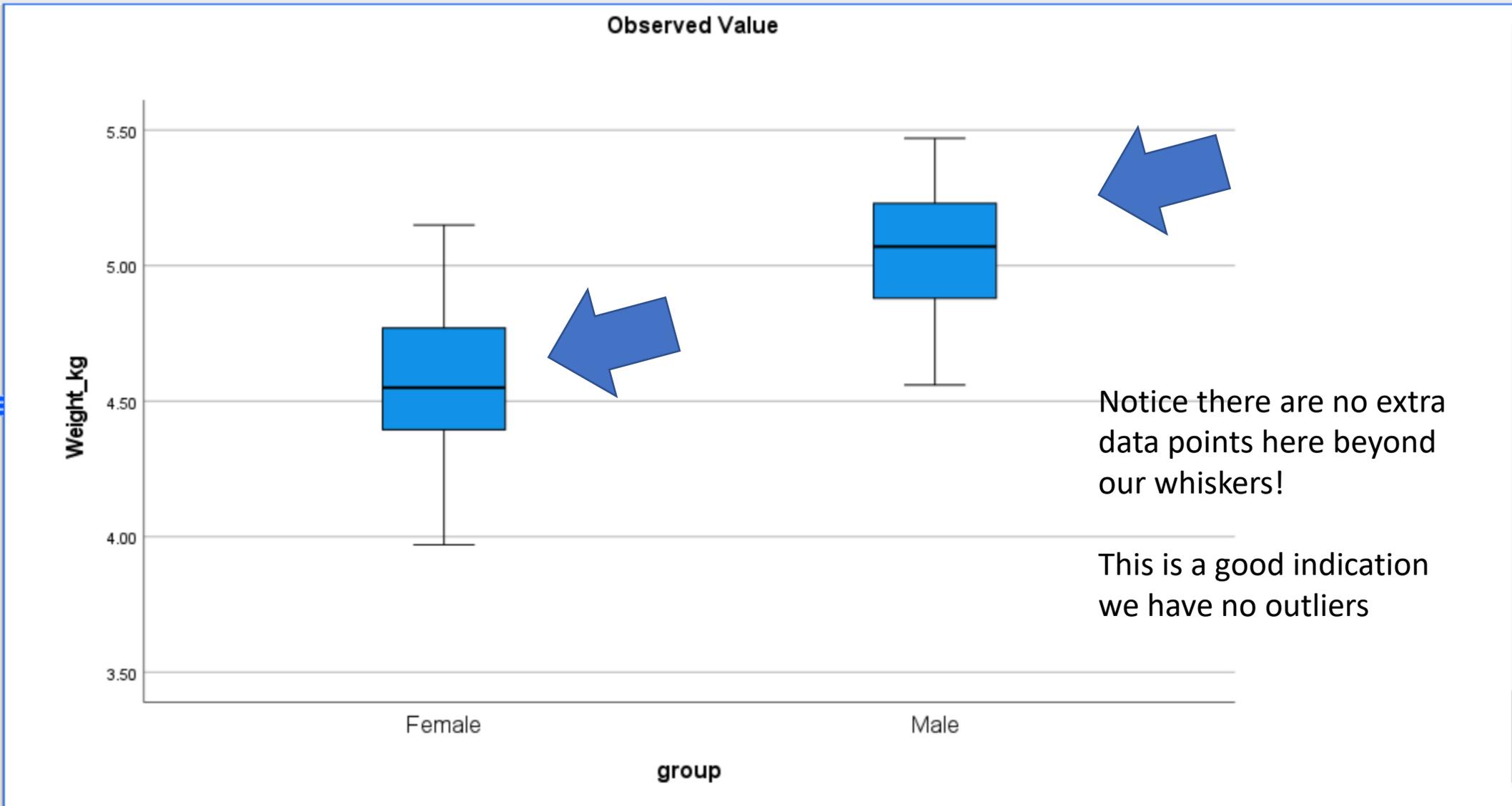
Double-click to activate



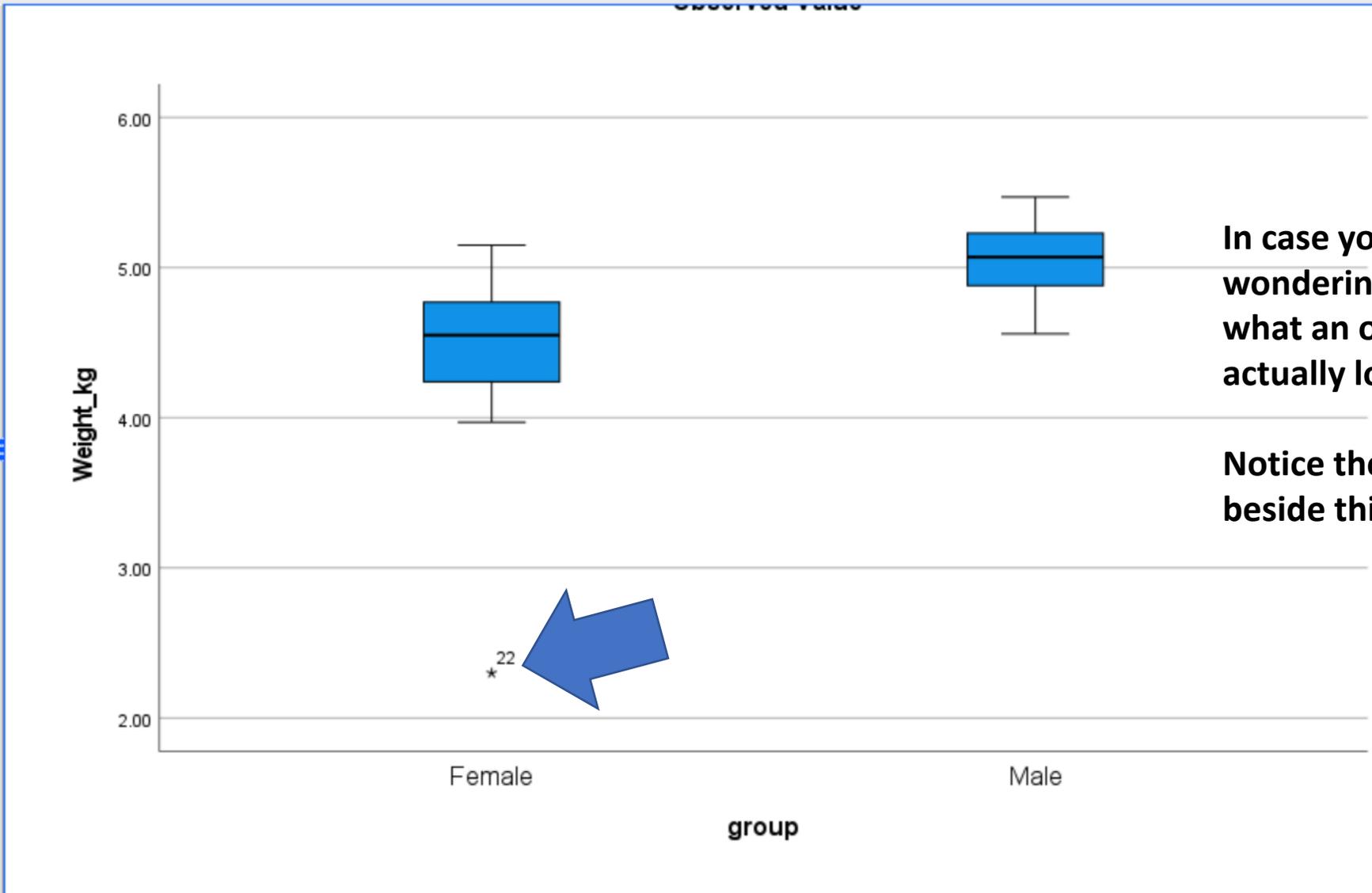
- Output
 - Explore
 - Title
 - Notes
 - Active Dataset
 - group
 - Title
 - Case Process
 - Descriptives
 - Tests of Norm
 - Weight_kg
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 - group
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- Output
 - Explore
 - Title
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 - Title
 - Case Process
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 - Extreme Value
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 - Weight_kg
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- Output
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 - group
 - group
 - Detrended
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 - group
 - group
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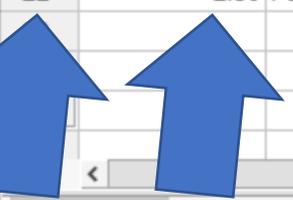
In case you were wondering... This is what an outlier might actually look like!

Notice the "22" beside this data point



25 : Visible: 2 of 2 Variables

	Weight_kg	group	var													
4	5.23	male														
5	5.44	Male														
6	5.13	Male														
7	4.88	Male														
8	4.79	Male														
9	4.98	Male														
10	5.01	Male														
11	4.56	Female														
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16	4.45	Female														
17	4.14	Female														
18	4.55	Female														
19	4.78	Female														
20	4.84	Female														
21	4.55	Female														
22	2.30	Female														



Data View Variable View

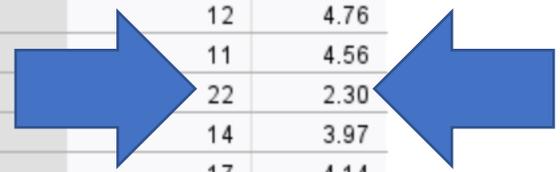


Output

- Explore
 - Title
 - Notes
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 - Case Process
 - Descriptives
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 - group
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 - Title
 - group
 - group
 - Detrended
 - Title
 - group
 - group
 - Boxplot

Extreme Values

group		Case Number	Value	
Weight_kg	Female	Highest 1	15	5.15
		2	20	4.84
		3	19	4.78
		4	12	4.76
		5	11	4.56
	Lowest	1	22	2.30
		2	14	3.97
		3	17	4.14
		4	13	4.34
		5	16	4.45
Male	Highest	1	2	5.47
		2	5	5.44
		3	4	5.23
		4	3	5.15
		5	6	5.13
	Lowest	1	1	4.56
		2	8	4.79
		3	7	4.88
		4	9	4.98
		5	10	5.01



Tests of Normality

group	Kolmogorov-Smirnov ^a			Shapiro-Wilk		
	Statistic	df	Sig.	Statistic	df	Sig.

Homogeneity of variance

- Is the distribution of the data around the mean approximately equal in each group?
- Good news, SPSS does a really nice thing for us here, and we don't have to test this assumption before we run the test...
- It will check for us during the test, and it gives us a solution if this assumption isn't met! 😊

Let's do this thing!

- Okay!
- Our data were collected independently!
- Our data appear normal!
- We don't have any outliers!
- Colin said we don't have to worry about the equality of variances yet...
- Can we please run a T Test!?!?!?

Yes!



1: Visible: 2 of 2 Variables

	Weight_kg	group	var												
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2	5.47	Male													
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1: Visible: 2 of 2 Variables

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18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
group

Estimate effect sizes

Buttons: Options..., Bootstrap..., OK, Paste, Reset, Cancel, Help



1: Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
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19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):

Weight_kg

group

Options...

Bootstrap...

Grouping Variable:

Define Groups...

Estimate effect sizes

OK Paste Reset Cancel Help



1: [Search bar] Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
4	5.23	Male													
5	5.44	Male													
6	5.13	Male													
7	4.88	Male													
8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
[Empty]

Estimate effect sizes

Options...
Bootstrap...

OK Paste Reset Cancel Help



1: Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
4	5.23	Male													
5	5.44	Male													
6	5.13	Male													
7	4.88	Male													
8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
group

Options...
Bootstrap...

Define Groups...

Estimate effect sizes

OK Paste Reset Cancel Help



1: [Search bar] Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
4	5.23	Male													
5	5.44	Male													
6	5.13	Male													
7	4.88	Male													
8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
group(??)

Estimate effect sizes

Buttons: Options..., Bootstrap..., Define Groups..., OK, Paste, Reset, Cancel, Help

Data View Variable View



1: Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
4	5.23	Male													
5	5.44	Male													
6	5.13	Male													
7	4.88	Male													
8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
group(??)

Estimate effect sizes

Options...
Bootstrap...

Define Groups...

OK Paste Reset Cancel Help





1: [Search bar] Visible: 2 of 2 Variables

	Weight_kg	group	var											
1	4.56	Male												
2	5.47	Male												
3	5.15	Male												
4	5.23	Male												
5	5.44	Male												
6	5.13	Male												
7	4.88	Male												
8	4.79	Male												
9	4.98	Male												
10	5.01	Male												
11	4.56	Female												
12	4.76	Female												
13	4.34	Female												
14	3.97	Female												
15	5.15	Female												
16	4.45	Female												
17	4.14	Female												
18	4.55	Female												
19	4.78	Female												
20	4.84	Female												
21	4.55	Female												
22														

Independent-Samples T Test

Test Variable(s):
Weight_kg

Options...
Bootstrap...

Define Groups

Group 1: Male
Group 2: Female

Continue Cancel

Define Groups...

Estimate effect sizes

OK Paste Reset Cancel Help

Data View Variable View



1: [Search bar] Visible: 2 of 2 Variables

	Weight_kg	group	var												
1	4.56	Male													
2	5.47	Male													
3	5.15	Male													
4	5.23	Male													
5	5.44	Male													
6	5.13	Male													
7	4.88	Male													
8	4.79	Male													
9	4.98	Male													
10	5.01	Male													
11	4.56	Female													
12	4.76	Female													
13	4.34	Female													
14	3.97	Female													
15	5.15	Female													
16	4.45	Female													
17	4.14	Female													
18	4.55	Female													
19	4.78	Female													
20	4.84	Female													
21	4.55	Female													
22															

Independent-Samples T Test

Test Variable(s):
Weight_kg

Grouping Variable:
group('Male' 'Female')

Estimate effect sizes

Buttons: Options..., Bootstrap..., Define Groups..., OK, Paste, Reset, Cancel, Help



- Test
- Title
- Notes
- Group St
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- Independ

Group Statistics

	group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg	Male	10	5.0640	.28254	.08935
	Female	11	4.5536	.33161	.09998

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means							
		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
	Equal variances not assumed			3.806	18.933	<.001	.001	.51036	.13409	.22965	.79108

Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

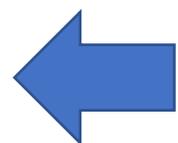
- a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.



- Test
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Group Statistics

group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg Male	10	5.0640	.28254	.08935
Weight_kg Female	11	4.5536	.33161	.09998



Descriptive stats by group:

- Number of cases – 10 males, and 11 females
- ★ Means – same as earlier
- ★ Standard deviation/errors for each mean

Independent Samples Test

Levene's Test for Equality of Variances				t-test for Equality of Means							
		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
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Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.



-Test
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Group Statistics

	group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg	Male	10	5.0640	.28254	.08935
	Female	11	4.5536	.33161	.09998

Independent Samples Test


 Levene's Test for Equality of Variances

t-test for Equality of Means

		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
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Independent Samples Effect Sizes

	Standardizer ^a	Point Estimate	95% Confidence Interval	
			Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	2.636
	Hedges' correction	.32226	1.584	2.531
	Glass's delta	.33161	1.539	2.600

a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

Levene's test for equality of variance:

- That other assumption: Is the variance around the mean for each group approximately equal?
- If this statistic is **not significant (0.05 or higher)**, then you can assume the variance is equal



- Test
- Title
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Group Statistics

	group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg	Male	10	5.0640	.28254	.08935
	Female	11	4.5536	.33161	.09998

Independent Samples Test

Levene's Test for Equality of Variances

t-test for Equality of Means

		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
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Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

- a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

SO: Since our Levene's test gave us a p value that was greater than (or equal to) 0.05, we can look at the first row of this table for our results 😊

Easy!

If this wasn't the case, you'd simply look at the second row: "Equal variance not assumed" welch's t-test



-Test
 Title
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Group Statistics

	group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg	Male	10	5.0640	.28254	.08935
	Female	11	4.5536	.33161	.09998

Independent Samples Test

		Levene's Test for Equality of Variances				t-test for Equality of Means				95% Confidence Interval of the Difference	
		F	Sig.	t	df	One-Sided p	Two-Sided p	Mean Difference	Std. Error Difference	Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
	Equal variances not assumed			3.806	18.933	<.001	.001	.51036	.13409	.22965	.79108

Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

a. The denominator used in estimating the effect sizes.

Cohen's d uses the pooled standard deviation.

Hedges' correction uses the pooled standard deviation, plus a correction factor.

Glass's delta uses the sample standard deviation of the control group.

- t is the computed test statistic (t statistic)
- df is the degrees of freedom

Title
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Group Statistics

group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg Male	10	5.0640	.28254	.08935
Weight_kg Female	11	4.5536	.33161	.09998

Independent Samples Test

Levene's Test for Equality of Variances



t-test for Equality of Means

		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
	Equal variances not assumed			3.806	18.933	<.001	.001	.51036	.13409	.22965	.79108

Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

- *Your p value!* Generally, this is the number people get the most excited about:
- Within the scientific community, we generally accept p values that are <0.05 as significant:
- In this case, our p value is <0.05, so we would say we can reject the null hypothesis that the mean weights in our 2 groups are equal



-Test
 Title
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 Group St
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 Independ

Group Statistics

group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg Male	10	5.0640	.28254	.08935
Weight_kg Female	11	4.5536	.33161	.09998

Independent Samples Test

Levene's Test for Equality of Variances

t-test for Equality of Means

		Levene's Test for Equality of Variances		t-test for Equality of Means				95% Confidence Interval of the Difference			
		F	Sig.	t	df	One-Sided p	Two-Sided p	Mean Difference	Std. Error Difference	Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
	Equal variances not assumed			3.806	18.933	<.001	.001	.51036	.13409	.22965	.79108

Independent Samples Effect Sizes

	Standardizer ^a	Point Estimate	95% Confidence Interval	
			Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	2.636
	Hedges' correction	.32226	1.584	2.531
	Glass's delta	.33161	1.539	2.600

a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

• *Mean Difference* is the difference between the sample means, i.e. $x_1 - x_2$; it also corresponds to the numerator of the test statistic for that test

• *Std. Error Difference* is the standard error of the mean difference estimate; it also corresponds to the denominator of the test statistic for that test



- Test
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Group Statistics

group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg Male	10	5.0640	.28254	.08935
Weight_kg Female	11	4.5536	.33161	.09998

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means							
		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
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Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

- a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

Effect size:

When using T-tests, we often also want to report the size of the effect we are reporting
 p values may tell you whether or not there is an effect, but...
 ...p values DO NOT tell us anything about effect size



-Test
 Title
 Notes
 Group St
 Independ
 Independ

Group Statistics

group	N	Mean	Std. Deviation	Std. Error Mean
Weight_kg Male	10	5.0640	.28254	.08935
Weight_kg Female	11	4.5536	.33161	.09998

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means							
		F	Sig.	t	df	Significance		Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
						One-Sided p	Two-Sided p			Lower	Upper
Weight_kg	Equal variances assumed	.059	.811	3.776	19	<.001	.001	.51036	.13516	.22747	.79325
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Independent Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Weight_kg	Cohen's d	.30934	1.650	.632	2.636
	Hedges' correction	.32226	1.584	.606	2.531
	Glass's delta	.33161	1.539	.431	2.600

- a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control group.

Effect size:

For effect size, we use Cohen's d: a **standardized** mean difference

General rule:

small effect ($d = 0.2$ and up)

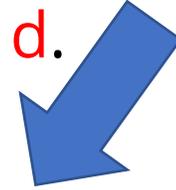
medium effect ($d = 0.5$ and up)

large effect ($d = 0.8$ and up)

So what am I reporting?!?!

- General formula:
- $t(\text{degrees of freedom}) = \text{the } t \text{ statistic}, p = p \text{ value}, d = \text{cohen's } d$.
- The mean weight of females in our sample was 4.55kg (SD = 0.33), whereas the mean weight in males was 5.06kg (SD = 0.28). An independent samples T Test showed that the difference was statistically significant, $t(19), 3.776, p = 0.001, d = 1.65$.

Take a look at this, and see where I got these numbers from our output!



Great! That wasn't so bad (I hope!)

- But what happens when we can't meet all the assumptions?
- Well, the assumption of equal variance is easy to deal with: welch's t-test

Independent Samples Test

		Levene's Test for Equality of Variances		t-test for Equality of Means ©						
		F	Sig.	t	df	Sig. (2-tailed)	Mean Difference	Std. Error Difference	95% Confidence Interval of the Difference	
									Lower	Upper
Mile time	Equal variances assumed	102.98	.000	13.475	390	.000	0:02:14	0:00:10	0:01:55	0:02:34
	Equal variances not assumed			15.047	315.846	.000	0:02:14	0:00:08	0:01:57	0:02:32

Image from: <https://libguides.library.kent.edu/spss/independentttest>

What about if our data isn't **normally distributed**?

- This is a much bigger issue.
- Lets take a look at an example...
- Load the data: “genderagg.xlsx”
 - Data containing the number of aggressive interactions started by male and female vervet monkeys over the span of a month



Visible: 3 of 3 Variables

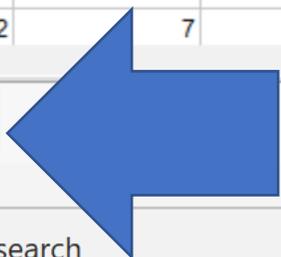
	Sex	Sex_num	aggress	var											
1	Female	1	19												
2	Female	1	18												
3	Female	1	9												
4	Female	1	17												
5	Female	1	8												
6	Female	1	7												
7	Female	1	16												
8	Female	1	19												
9	Female	1	20												
10	Female	1	9												
11	Female	1	11												
12	Female	1	18												
13	Male	2	16												
14	Male	2	5												
15	Male	2	15												
16	Male	2	2												
17	Male	2	14												
18	Male	2	15												
19	Male	2	4												
20	Male	2	7												
21	Male	2	15												
22	Male	2	6												
23	Male	2	7												

Take some time to look at the **average number of aggressive interactions** by males and females

Please notice here that Sex_num represents males and females...

This is important!

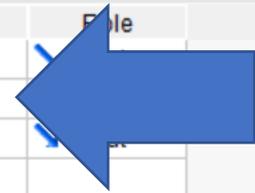
Check out your variable view



Data View Variable View



	Name	Type	Width	Decimals	Label	Values	Missing	Columns	Align	Measure	File
1	Sex	String	6	0		None	None	6	Left	Nominal	
2	Sex_num	Numeric	2	0		None	None	12	Right	Nominal	
3	aggress	Numeric	3	0		None	None	12	Right	Nominal	
4											
5											
6											
7											
8											
9											
10											
11											
12											
13											
14											
15											
16											
17											
18											
19											
20											
21											
22											
23											
24											
25											



Today, SPSS was on our size, and it automatically realized that the 1's and 2's for Sex_num were categories...

BUT



	Name	Type	Width	Decimals	Label	Values	Missing	Columns	Align	Measure	Role
1	Sex	String	6	0		None	None	6	Left	Nominal	Input
2	Sex_num	Numeric	2	0		None	None	12	Right	Nominal	
3	aggress	Numeric	3	0		None	None	12	Right	Nominal	
4											
5											
6											
7											
8											
9											
10											
11											
12											
13											
14											
15											
16											
17											
18											
19											
20											
21											
22											
23											
24											
25											



It got the measurement type wrong for 'aggress'!

This is a COUNT of aggressive interactions, and it is continuous



	Name	Type	Width	Decimals	Label	Values	Missing	Columns	Align	Measure	Role
1	Sex	String	6	0		None	None	6	Left	Nominal	Input
2	Sex_num	Numeric	2	0		None	None	12	Right	Nominal	Input
3	aggress	Numeric	3	0		None	None	12	Right	Nominal	Input
4										Scale	
5										Ordinal	
6										Nominal	
7											
8											
9											
10											
11											
12											
13											
14											
15											
16											
17											
18											
19											
20											
21											
22											
23											
24											
25											

Measure dropdown menu:

- Nominal
- Scale
- Ordinal
- Nominal

Click that box and be sure to change it to a 'Scale' variable!



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var											
1	Female	1	19												
2	Female	1	18												
3	Female	1	9												
4	Female	1	17												
5	Female	1	8												
6	Female	1	7												
7	Female	1	16												
8	Female	1	19												
9	Female	1	20												
10	Female	1	9												
11	Female	1	11												
12	Female	1	18												
13	Male	2	16												
14	Male	2	5												
15	Male	2	15												
16	Male	2	2												
17	Male	2	14												
18	Male	2	15												
19	Male	2	4												
20	Male	2	7												
21	Male	2	15												
22	Male	2	6												
23	Male	2	7												

You should try to make a **boxplot** to get a feel for the data

Go for it! Help each other out!!!

While you're at it, do you remember how to check out the **mean?**



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var											
1	Female	1	19												
2	Female	1	18												
3	Female	1	9												
4	Female	1	17												
5	Female	1	8												
6	Female	1	7												
7	Female	1	16												
8	Female	1	19												
9	Female	1	20												
10	Female	1	9												
11	Female	1	11												
12	Female	1	18												
13	Male	2	16												
14	Male	2	5												
15	Male	2	15												
16	Male	2	2												
17	Male	2	14												
18	Male	2	15												
19	Male	2	4												
20	Male	2	7												
21	Male	2	15												
22	Male	2	6												
23	Male	2	7												

Now: This looks like a perfect data set to analyse using a paired T Test, but is it normally distributed?

Let's check!

Try it out!



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var											
1	Female	1	19												
2	Female	1	18												
3	Female	1	9												
4	Female	1	17												
5	Female	1	8												
6	Female	1	7												
7	Female	1	16												
8	Female	1	19												
9	Female	1	20												
10	Female	1	9												
11	Female	1	11												
12	Female	1	18												
13	Male	2	16												
14	Male	2	5												
15	Male	2	15												
16	Male	2	2												
17	Male	2	14												
18	Male	2	15												
19	Male	2	4												
20	Male	2	7												
21	Male	2	15												
22	Male	2	6												
23	Male	2	7												

As this data is not normally distributed, we cannot use a T Test

A commonly used non-parametric equivalent of the independent samples T Test is the **Mann-Whitney U test**

the **Mann-Whitney U test** drops the assumption of normality

Mann–Whitney U test

- A **Mann–Whitney U test** does not compare means of 2 groups
- It instead RANKS the values of the dependant values from lowest to highest

Sex	Sex_num	aggress
Female	1	19
Female	1	18
Female	1	9
Female	1	17
Female	1	8
Female	1	7
Female	1	16
Female	1	19
Female	1	20
Female	1	9
Female	1	11
Female	1	18
Male	2	16
Male	2	5
Male	2	15
Male	2	2
Male	2	14
Male	2	15
Male	2	4
Male	2	7
Male	2	15
Male	2	6
Male	2	7
Male	2	14

Mann–Whitney U test

- It then compares the **mean ranks** of the two groups:
- Null hypothesis: The two samples come from the same population
- I.e., the probability is 50% that a randomly drawn member of the first population will exceed a member of the second population
- I.e., does one group have higher ranking values than the other group overall?

1	Sex	Sex_num	aggress	Rank
2	Male	2	2	1
3	Male	2	4	2
4	Male	2	5	3
5	Male	2	6	4
6	Female	1	7	6
7	Male	2	7	6
8	Male	2	7	6
9	Female	1	8	8
10	Female	1	9	9.5
11	Female	1	9	9.5
12	Female	1	11	11
13	Male	2	14	12.5
14	Male	2	14	12.5
15	Male	2	15	15
16	Male	2	15	15
17	Male	2	15	15
18	Female	1	16	17.5
19	Male	2	16	17.5
20	Female	1	17	19
21	Female	1	18	20.5
22	Female	1	18	20.5
23	Female	1	19	22.5
24	Female	1	19	22.5
25	Female	1	20	24



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var											
1	Male	2	2												
2	Male	2	4												
3	Male	2	5												
4	Male	2	6												
5	Female	1	7												
6	Male	2	7												
7	Male	2	7												
8	Female	1	8												
9	Female	1	9												
10	Female	1	9												
11	Female	1	11												
12	Male	2	14												
13	Male	2	14												
14	Male	2	15												
15	Male	2	15												
16	Male	2	15												
17	Female	1	16												
18	Male	2	16												
19	Female	1	17												
20	Female	1	18												
21	Female	1	18												
22	Female	1	19												
23	Female	1	19												

Data View Variable View



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Two-Independent-Samples Tests

Test Variable List:

- Sex_num
- aggress

Grouping Variable:

Test Type

- Mann-Whitney U
- Kolmogorov-Smirnov Z
- Moses extreme reactions
- Wald-Wolfowitz runs

Buttons: OK, Paste, Reset, Cancel, Help



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Two-Independent-Samples Tests

Test Variable List: aggress

Grouping Variable:

Test Type

- Mann-Whitney U
- Kolmogorov-Smirnov Z
- Moses extreme reactions
- Wald-Wolfowitz runs

Buttons: OK, Paste, Reset, Cancel, Help

Pop 'aggress' in as the Test Variable

Pop 'Sex_num' in as the grouping variable

Notice you don't have a choice of using 'Sex'? In SPSS, this test needs to see numbers here... Even if it is coding for a categorical variable



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Two-Independent-Samples Tests

Test Variable List:
aggress

Exact...
Options...

Two Independent Samples: D...

Group 1:
Group 2:

Continue Cancel Help

Test Type

Mann-Whitney U Kolmogorov-Smirnov Z
 Moses extreme reactions Wald-Wolfowitz runs

OK Paste Reset Cancel Help



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Two-Independent-Samples Tests

Test Variable List:
aggress

Exact...
Options...

Two Independent Samples: D...
Group 1: 1
Group 2: 2
Continue Cancel Help

Test Type
 Mann-Whitney U Kolmogorov-Smirnov Z
 Moses extreme reactions Wald-Wolfowitz runs

OK Paste Reset Cancel Help



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Two-Independent-Samples Tests

Test Variable List:
aggress

Grouping Variable:
Sex_num(1 2)

Test Type:
 Mann-Whitney U
 Kolmogorov-Smirnov Z
 Moses extreme reactions
 Wald-Wolfowitz runs

Buttons: OK, Paste, Reset, Cancel, Help, Exact..., Options..., Define Groups...



*Output1 [Document1] - IBM SPSS Statistics Viewer

File Edit View Data Transform Insert Format Analyze Graphs Utilities Extensions Window Help

Output

- Output
 - NPar Tests
 - Title
 - Notes
 - Active Dataset
 - Mann-Whitney Test
 - Title
 - Ranks
 - Test Statistics

NPar Tests

[DataSet2]

Mann-Whitney Test

Ranks

	Sex_num	N	Mean Rank	Sum of Ranks
aggress	1	12	15.88	190.50
	2	12	9.13	109.50
	Total	24		

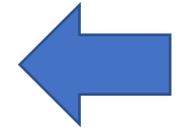
Test Statistics^a

	aggress
Mann-Whitney U	31.500
Wilcoxon W	109.500
Z	-2.345
Asymp. Sig. (2-tailed)	.019
Exact Sig. [2*(1-tailed Sig.)]	.017 ^b

a. Grouping Variable: Sex_num
b. Not corrected for ties.

Our first bit of output is the **ranks** table.

Here, we can see that Sex_num 1 (females) had a higher mean rank than Sex_num 2 (males)



Importantly: If sex had no effect on the number of aggressive interactions an individual took part in, we would expect the mean ranks of males and females to be quite similar

Clipboard: Paste, Copy, Cut, Undo, Redo

Font: Calibri, 11, Bold, Italic, Underline, Text Color, Background Color, Font Color

Alignment: Center, Left, Right, Merge & Center, Wrap Text

Number: General, Percentage, Decimals

Styles: Conditional Formatting, Format as Table, Cell Styles

Cells: Insert, Delete, Format

Editing: Sort & Filter, Find & Select

Analysis: Analyze Data

M7

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
1	Male	2	2	1			Female	1	7	6							
2	Male	2	4	2			Female	1	8	8							
3	Male	2	5	3			Female	1	9	9.5							
4	Male	2	6	4			Female	1	9	9.5							
5	Male	2	7	6			Female	1	11	11							
6	Male	2	7	6			Female	1	16	17.5							
7	Male	2	14	12.5			Female	1	17	19							
8	Male	2	14	12.5			Female	1	18	20.5							
9	Male	2	15	15			Female	1	18	20.5							
10	Male	2	15	15			Female	1	19	22.5							
11	Male	2	15	15			Female	1	19	22.5							
12	Male	2	16	17.5			Female	1	20	24							
13																	
14																	
15			Average Rank Males	9.125					Average Rank females	15.875							
16																	
17			Sum or Ranks Males	109.5					Sum of Ranks Females	190.5							
18																	
19																	
20																	
21																	

*Output1 [Document1] - IBM SPSS Statistics Viewer

File Edit View Data Transform Insert Format Analyze Graphs Utilities Extensions Window Help

Output

- Output
 - NPar Tests
 - Title
 - Notes
 - Active Dataset
 - Mann-Whitney Test
 - Title
 - Ranks
 - Test Statistics

NPar Tests

[DataSet2]

Mann-Whitney Test

Ranks

	Sex_num	N	Mean Rank	Sum of Ranks
aggress	1	12	15.88	190.50
	2	12	9.13	109.50
Total		24		

Test Statistics^a

	aggress
Mann-Whitney U	31.500
Wilcoxon W	109.500
Z	-2.345
Asymp. Sig. (2-tailed)	.019
Exact Sig. [2*(1-tailed Sig.)]	.017 ^b

a. Grouping Variable: Sex_num
b. Not corrected for ties.



Mann-Whitney U and **Wilcoxon W** are test statistics; they summarize the difference in mean rank.

We prefer reporting Exact Sig. (2-tailed): the p-value corrected for ties.

Importantly, this p-value is lower than 0.05

This means we can say that there was a significant difference in the mean ranks of our 2 samples!

This suggests a difference between the 2 groups

Reporting

- General formula:
- $U =$ (value of Mann-Whitney U test statistic), $p =$ (p value)
 - Also, you should generally report the medians, and mean rank values for your two groups.
 - A Mann-Whitney test indicated that the number of aggressive interactions that an individual engaged in per day was greater for females (Mdn = ★ , MeanRk = 15.88) than it was for males (Mdn = ★ , MeanRk = 9.13), $U = 31.5$, $p = 0.019$.

- ★
- Importantly, all this info is available on your SPSS output, except for **the medians...** I can show you where that comes from super quick!
- ★



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Means

Dependent List: aggress

Sex_num

Layer 1 of 1

Previous Next

Layer 1 of 1

Sex

Options... Style... Bootstrap...

OK Paste Reset Cancel Help



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Means

Dependent List: aggress

Sex_num

Layer 1 of 1

Previous Next

Layer 1 of 1

Sex

Options... Style... Bootstrap...

OK Paste Reset Cancel Help



	Sex	Sex_num	aggress	var	var
1	Male	2	2		
2	Male	2	4		
3	Male	2	5		
4	Male	2	6		
5	Female	1	7		
6	Male	2	7		
7	Male	2	7		
8	Female	1	8		
9	Female	1	9		
10	Female	1	9		
11	Female	1	11		
12	Male	2	14		
13	Male	2	14		
14	Male	2	15		
15	Male	2	15		
16	Male	2	15		
17	Female	1	16		
18	Male	2	16		
19	Female	1	17		
20	Female	1	18		
21	Female	1	18		
22	Female	1	19		
23	Female	1	19		

Visible: 3 of 3 Variables

Statistics:

- Median
- Std. Error of Mean
- Sum
- Minimum
- Maximum
- Range
- First
- Last
- Variance
- Kurtosis
- Std. Error of Kurtosis
- Skewness
- Std. Error of Skewness
- Harmonic Mean

Cell Statistics:

- Mean
- Number of Cases
- Standard Deviation

Statistics for First Layer

- Anova table and eta
- Test for linearity

Continue Cancel Help



	Sex	Sex_num	aggress	var	var
1	Male	2	2		
2	Male	2	4		
3	Male	2	5		
4	Male	2	6		
5	Female	1	7		
6	Male	2	7		
7	Male	2	7		
8	Female	1	8		
9	Female	1	9		
10	Female	1	9		
11	Female	1	11		
12	Male	2	14		
13	Male	2	14		
14	Male	2	15		
15	Male	2	15		
16	Male	2	15		
17	Female	1	16		
18	Male	2	16		
19	Female	1	17		
20	Female	1	18		
21	Female	1	18		
22	Female	1	19		
23	Female	1	19		

Visible: 3 of 3 Variables

Statistics:

- Std. Error of Mean
- Sum
- Minimum
- Maximum
- Range
- First
- Last
- Variance
- Kurtosis
- Std. Error of Kurtosis
- Skewness
- Std. Error of Skewness
- Harmonic Mean
- Geometric Mean

Cell Statistics:

- Mean
- Number of Cases
- Standard Deviation
- Median

Statistics for First Layer

- Anova table and eta
- Test for linearity

Buttons: Continue, Cancel, Help



Visible: 3 of 3 Variables

	Sex	Sex_num	aggress	var										
1	Male	2	2											
2	Male	2	4											
3	Male	2	5											
4	Male	2	6											
5	Female	1	7											
6	Male	2	7											
7	Male	2	7											
8	Female	1	8											
9	Female	1	9											
10	Female	1	9											
11	Female	1	11											
12	Male	2	14											
13	Male	2	14											
14	Male	2	15											
15	Male	2	15											
16	Male	2	15											
17	Female	1	16											
18	Male	2	16											
19	Female	1	17											
20	Female	1	18											
21	Female	1	18											
22	Female	1	19											
23	Female	1	19											

Means

Dependent List: aggress

Sex_num

Layer 1 of 1

Previous Next

Layer 1 of 1

Sex

Options... Style... Bootstrap...

OK Paste Reset Cancel Help



- Output
 - Means
 - Title
 - Notes
 - Case Processing Summary
 - Report

➔ Means

Case Processing Summary

	Included		Cases Excluded		Total	
	N	Percent	N	Percent	N	Percent
aggress * Sex	24	100.0%	0	0.0%	24	100.0%

Report

aggress

Sex	Mean	N	Std. Deviation	Median
Female	14.25	12	4.993	16.50
Male	10.00	12	5.240	10.50
Total	12.13	24	5.456	14.00

Reporting

- General formula:
- $U = (\text{value of Mann-Whitney } U \text{ test statistic}), p = (\text{p value})$
 - Also, you should generally report the medians, and mean rank values for your two groups.
- Let's add in the median values!
- A Mann-Whitney test indicated that the number of aggressive interactions that an individual engaged in per day was greater for females (Mdn = 16.5, MeanRk = 15.88) than it was for males (Mdn = 10.5, MeanRk = 9.13), $U = 31.5, p = 0.019$.

Paired groups

- In our first 2 analyses, we were comparing 2 groups
- In those tests, each group was made up of different individuals
- The males in group one were different individuals than the females in group 2!

- But what if we measure the SAME individuals twice?
 - i.e., measure the animals once before an experimental treatment
 - Then measure the SAME ANIMALS after the treatment.

- This requires different tests!

Paired samples T Test

- **Paired Samples T Test** compares the means of related group of samples.
 - It is used when you have 2 values measured **from the same individual** – each measurement representing a different time, treatment, etc...
 - Our measurements are therefore “paired”
- Examples:
 - You might want to compare the average weight of 20 mice **before**, and **after** a treatment
 - The data contains 20 sets of values before treatment, and 20 sets of values after treatment - the weight of each mouse was weighed twice (once before treatment, and once after treatment)
 - A measurement taken at two different times (e.g., Weight of animals in the fall compared to the spring)
 - The data contains 15 sets of values from the fall, and 15 from the spring – each animal was captured and weighed in the fall, then **re-captured** and weighed **again** in the winter

The data

- Let's open the data "mice2.xlsx"

	id	before	after	var												
1	1	187.2	429.5													
2	2	194.2	404.4													
3	3	231.7	405.6													
4	4	200.5	397.2													
5	5	201.7	377.9													
6	6	235.0	445.8													
7	7	208.7	408.4													
8	8	172.4	337.0													
9	9	184.6	414.3													
10	10	189.6	380.3													
11																
12																
13																
14																
15																
16																
17																
18																
19																
20																
21																
22																
23																

- You are made aware that a new shopping centre with a large food court is moving into your town
- You are concerned, as your favourite species, the **brown rat**, live in your town, and you are afraid that the access to food court garbage will negatively affect their diet, and weight.
- You set out to quickly collect weight data before the shopping food court is built to get some baseline data!

	id	before	after	var												
1	1	187.2	429.5													
2	2	194.2	404.4													
3	3	231.7	405.6													
4	4	200.5	397.2													
5	5	201.7	377.9													
6	6	235.0	445.8													
7	7	208.7	408.4													
8	8	172.4	337.0													
9	9	184.6	414.3													
10	10	189.6	380.3													
11																
12																
13																
14																
15																
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22																
23																

- Our data represent the weights of 10 individual brown rats
- Individual IDs are indicated by the numbers (1 through 10) under the 'id' column
- The 'before' column represents weight measurements of these individuals **before the food court was operational**
- The individual rats were caught, marked, weighed, and released back into the 'wilds' of your town
- The column marked 'after' represents the weight of those **SAME RATS**, who were recaptured 5 months **AFTER the food court was finished and opened**



13 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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23															

Before we move on,
we should really
explore our new
data!

Data View Variable View



13 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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23															



6 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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21															
22															
23															

Means

Dependent List:

id

after

before

Layer 1 of 1

Options...

Style...

Bootstrap...

Previous

Next

Layer 1 of 1

OK Paste Reset Cancel Help



6 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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21															
22															
23															

Means

Dependent List:

id

after

before

Layer 1 of 1

Previous Next

Layer 1 of 1

OK Paste Reset Cancel Help

Options... Style... Bootstrap...



6 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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18															
19															
20															
21															
22															
23															

Means

Dependent List:
after
before

id

Layer 1 of 1
Previous Next

Layer 1 of 1

OK Paste Reset Cancel Help

Options...
Style...
Bootstrap...



	id	after	before	var	var
1	1	429.5	187.2		
2	2	404.4	194.2		
3	3	405.6	231.7		
4	4	397.2	200.5		
5	5	377.9	201.7		
6	6	445.8	235.0		
7	7	408.4	208.7		
8	8	337.0	172.4		
9	9	414.3	184.6		
10	10	380.3	189.6		
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21					
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23					

Visible: 3 of 3 Variables

Statistics:

- Median
- Grouped Median
- Std. Error of Mean
- Sum
- Minimum
- Maximum
- Range
- First
- Last
- Variance
- Kurtosis
- Std. Error of Kurtosis
- Skewness
- Std. Error of Skewness

Cell Statistics:

- Mean
- Number of Cases
- Standard Deviation

Statistics for First Layer

- Anova table and eta
- Test for linearity

Continue Cancel Help



6 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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19															
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21															
22															
23															

Means

Dependent List:

- after
- before

id

Layer 1 of 1

Previous Next

Layer 1 of 1

OK Paste Reset Cancel Help

Options... Style... Bootstrap...

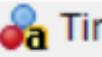
Data structure – Wide format

- You will notice that last time we looked at our average values this way, we only had to move one variable over... Why is there 2 this time?
- This is because our data is in a different format this time:
- Our current data is in **wide format**:
 - **Each row of data** represents a single individual
 - The treatment (the independent variable) are now column headers!

	 id	 after	 before
1	1	429.5	187.2
2	2	404.4	194.2
3	3	405.6	231.7
4	4	397.2	200.5
5	5	377.9	201.7
6	6	445.8	235.0
7	7	408.4	208.7
8	8	337.0	172.4
9	9	414.3	184.6
10	10	380.3	189.6
11			
12			

Data structure – Long format

- You will notice that last time we looked at our average values this way, we only had to move one variable over... Why is there 2 this time?
- This is because our data is in a different format this time:
- Our current data is in wide format: **Each row of data** represents a single individual
- Our independent variable is listed for each row

	 id	 Time	 Weight
1	1	before	187.2
2	1	after	429.5
3	2	before	194.2
4	2	after	404.4
5	3	before	231.7
6	3	after	405.6
7	4	before	200.5
8	4	after	397.2
9	5	before	201.7
10	5	after	377.9
11	6	before	235.0
12	6	after	445.8
13	7	before	208.7
14	7	after	408.4
15	8	before	172.4
16	8	after	337.0
17	9	before	184.6
18	9	after	414.3
19	10	before	189.6
20	10	after	380.3



6 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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21															
22															
23															

Means

Dependent List:
after
before

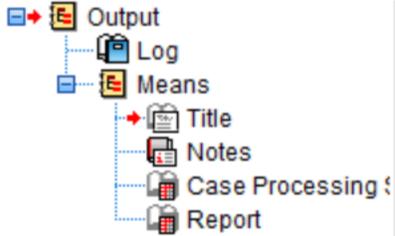
id

Layer 1 of 1
Previous Next

Layer 1 of 1

OK Paste Reset Cancel Help

Options...
Style...
Bootstrap...



```
MEANS TABLES=after before  
/CELLS=MEAN COUNT STDDEV.
```

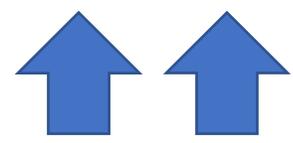
➔ Means

Case Processing Summary

	Included		Cases Excluded		Total	
	N	Percent	N	Percent	N	Percent
after	10	100.0%	0	0.0%	10	100.0%
before	10	100.0%	0	0.0%	10	100.0%

Report

	after	before
Mean	400.040	200.560
N	10	10
Std. Deviation	30.0869	20.0282



Notice: This time around, our two categories (before and after) are in 2 columns... but they were in rows before, right?!

Again, this is because our data are in **wide format** as opposed to **long format**



9 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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Our next course of action should be to graph our data

We could try to use the same procedure as we did before, but here is where the issue of **long vs wide** formatted data becomes a real problem...

In order to make the boxplot we need, we need to covert from our current **wide format** into a **long format!**



9 : Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
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21															
22															
23															



	id	before	after
1	1	187.2	429.5
2	2	194.2	404.4
3	3	231.7	405.6
4	4	200.5	397.2
5	5	201.7	377.9
6	6	235.0	445.8
7	7	208.7	408.4
8	8	172.4	337.0
9	9	184.6	414.3
10	10	189.6	380.3
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Variables to Cases: Create One Index Variable

You have chosen to create one index variable. The variable's values can be sequential numbers or the names of variables in a group.

In the table you can specify the name and label for the index variable.

What kind of index values?

Sequential numbers

Index Values: 1, 2

Variable names

Index Values: before, after

Edit the Index Variable Name and Label:

	Name	Label	Levels	Index Values
1	Index1		2	before, after



You can also rename the column header that will contain the values of 'before' and 'after'...

Lets call that variable 'Time'



	id	before	after
1	1	187.2	429.5
2	2	194.2	404.4
3	3	231.7	405.6
4	4	200.5	397.2
5	5	201.7	377.9
6	6	235.0	445.8
7	7	208.7	408.4
8	8	172.4	337.0
9	9	184.6	414.3
10	10	189.6	380.3
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18			
19			
20			
21			
22			
23			

Variables to Cases: Create One Index Variable

You have chosen to create one index variable. The variable's values can be sequential numbers or the names of variables in a group.

In the table you can specify the name and label for the index variable.

What kind of index values?

Sequential numbers

Index Values: 1, 2

Variable names

Index Values: before, after

Edit the Index Variable Name and Label:

	Name	Label	Levels	Index Values
1	Time		2	before, after



At this point, there are more options, but we are done for today.

Click 'Finish'

	id	Time	Weight	var												
1	1	before	187.2													
2	1	after	429.5													
3	2	before	194.2													
4	2	after	404.4													
5	3	before	231.7													
6	3	after	405.6													
7	4	before	200.5													
8	4	after	397.2													
9	5	before	201.7													
10	5	after	377.9													
11	6	before	235.0													
12	6	after	445.8													
13	7	before	208.7													
14	7	after	408.4													
15	8	before	172.4													
16	8	after	337.0													
17	9	before	184.6													
18	9	after	414.3													
19	10	before	189.6													
20	10	after	380.3													
21																
22																
23																

And our data is now in **long format:**

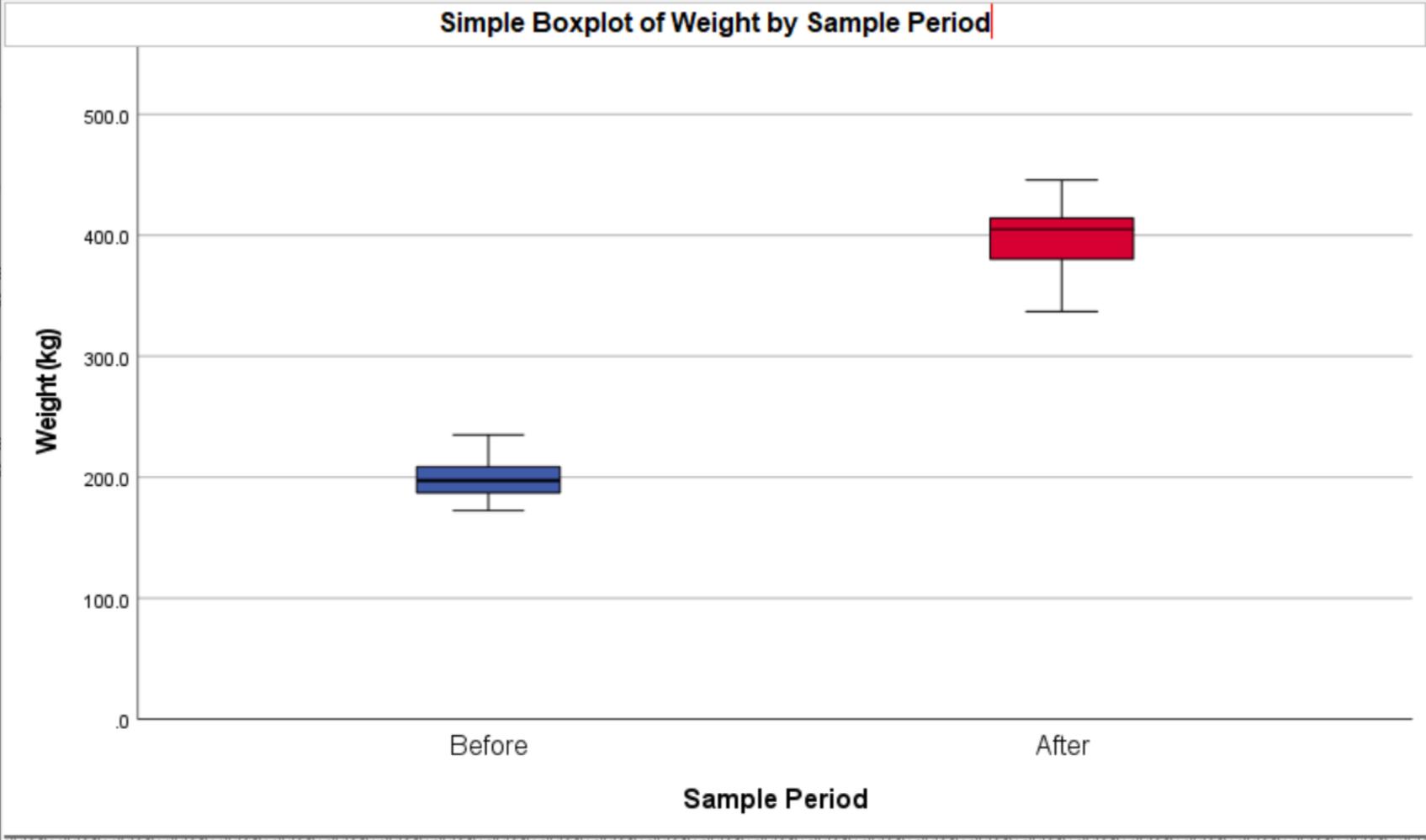
This looks much more like what we had for our two-sampled T Test!

	id	Time	Weight	var												
1	1	before	187.2													
2	1	after	429.5													
3	2	before	194.2													
4	2	after	404.4													
5	3	before	231.7													
6	3	after	405.6													
7	4	before	200.5													
8	4	after	397.2													
9	5	before	201.7													
10	5	after	377.9													
11	6	before	235.0													
12	6	after	445.8													
13	7	before	208.7													
14	7	after	408.4													
15	8	before	172.4													
16	8	after	337.0													
17	9	before	184.6													
18	9	after	414.3													
19	10	before	189.6													
20	10	after	380.3													
21																
22																
23																

Now you should be able to make you boxplot! Give it a try!



- Output
 - Log
 - Means
 - Title
 - Notes
 - Case Proces
 - Report
 - Log
 - Variables to Case
 - Title
 - Notes
 - Generated Va
 - Processing S
 - Log
 - Variables to Case
 - Title
 - Notes
 - Generated Va
 - Processing S
 - Log
 - GGraph
 - Title
 - Notes
 - Graph





Visible: 3 of 3 Variables

	id	Time	height	var											
1	1	before	187.2												
2	1	after	429.5												
3	2	before	194.2												
4	2	after	404.4												
5	3	before	231.7												
6	3	after	405.6												
7	4	before	200.5												
8	4	after	397.2												
9	5	before	201.7												
10	5	after	377.9												
11	6	before	235.0												
12	6	after	445.8												
13	7	before	208.7												
14	7	after	408.4												
15	8	before	172.4												
16	8	after	337.0												
17	9	before	184.6												
18	9	after	414.3												
19	10	before	189.6												
20	10	after	380.3												
21															
22															
23															



Well great job!

So, we are ready to move forward with our Paired T Test... but guess what:

That analysis wants our data to be back in **wide format**

We Have to covert it back!?!?!?

This procedure is much shorter actually



Visible: 3 of 3 Variables

	id	Time	Weight	var												
1		1 before	187.2													
2		1 after	429.5													
3		2 before	194.2													
4		2 after	404.4													
5		3 before	231.7													
6		3 after	405.6													
7		4 before	200.5													
8		4 after	397.2													
9		5 before	201.7													
10		5 after	377.9													
11		6 before	235.0													
12		6 after	445.8													
13		7 before	208.7													
14		7 after	408.4													
15		8 before	172.4													
16		8 after	337.0													
17		9 before	184.6													
18		9 after	414.3													
19		10 before	189.6													
20		10 after	380.3													
21																
22																
23																

IBM SPSS Statistics

Save changes to the following dataset?

Dataset Name: DataSet4
File Name: Untitled5

Yes No Cancel



	id	Time	Weight
1	1	before	187.2
2	1	after	429.5
3	2	before	194.2
4	2	after	404.4
5	3	before	231.7
6	3	after	405.6
7	4	before	200.5
8	4	after	397.2
9	5	before	201.7
10	5	after	377.9
11	6	before	235.0
12	6	after	445.8
13	7	before	208.7
14	7	after	408.4
15	8	before	172.4
16	8	after	337.0
17	9	before	184.6
18	9	after	414.3
19	10	before	189.6
20	10	after	380.3
21			
22			
23			

Cases to Variables: Sorting Data

The variables that you used to identify case groups in the current file need to be sorted before the file can be restructured. If you are not sure about your data, select "Yes".

Sort the current data?

2	1	3	.006
3	1	1	.010
1	1	1	.003
2	1	1	.008
2	1	2	.007
1	1	2	.004
1	1	3	.002

1	1	1	.003
1	1	2	.004
1	1	3	.002
2	1	1	.008
2	1	2	.007
2	1	3	.006
3	1	1	.010

Yes - data will be sorted by the Identifier and Index variables

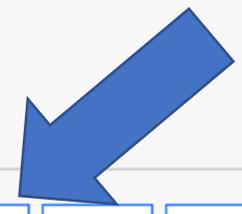
2	1	3	.006
3	1	1	.010
1	1	1	.003
2	1	1	.008
2	1	2	.007
1	1	2	.004
1	1	3	.002

1	1	1	.003
1	1	2	.004
1	1	3	.002
2	1	1	.008
2	1	2	.007
2	1	3	.006
3	1	1	.010

No - use the data as currently sorted



Keep default of 'yes', then select 'Finish'





Visible: 3 of 3 Variables

	id	after	before	var											
1	1	429.5	187.2												
2	2	404.4	194.2												
3	3	405.6	231.7												
4	4	397.2	200.5												
5	5	377.9	201.7												
6	6	445.8	235.0												
7	7	408.4	208.7												
8	8	337.0	172.4												
9	9	414.3	184.6												
10	10	380.3	189.6												
11															
12															
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14															
15															
16															
17															
18															
19															
20															
21															
22															
23															

And we are back to **wide format!**

We can start looking at our paired samples T Test...

Almost

Assumptions

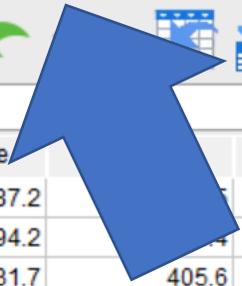
- Similar to the two sample T Test, we must check that our data meet some assumptions:
- The two groups are 'paired'.
 - In our example, this is the case as the data were collected measuring the **same individual rats at 2 different time periods.**
- No significant outliers in **the difference between the two related groups**
- Normality: **the difference between the two related groups** follow a normal distribution



Visible: 3 of 3 Variables

	id	before	after	var											
1	1	187.2	429.5												
2	2	194.2	404.4												
3	3	231.7	405.6												
4	4	200.5	397.2												
5	5	201.7	377.9												
6	6	235.0	445.8												
7	7	208.7	408.4												
8	8	172.4	337.0												
9	9	184.6	414.3												
10	10	189.6	380.3												
11															
12															
13															
14															
15															
16															
17															
18															
19															
20															
21															
22															
23															

SO: We need to look at the distribution of **the difference between the two groups**. We are interested in:
1) The distribution: is it normal?
2) Are there outliers?



Visible: 3 of 3 Variables

	id	before	var	var	var	var	var	var	var	var	var	var	var	var
1	1	187.2												
2	2	194.2												
3	3	231.7	405.6											
4	4	200.5	397.2											
5	5	201.7	377.9											
6	6	235.0	445.8											
7	7	208.7	408.4											
8	8	172.4	337.0											
9	9	184.6	414.3											
10	10	189.6	380.3											
11														
12														
13														
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16														
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22														
23														

Step one:
let's calculate **the difference between the two groups!**



- Compute Variable...
- Programmability Transformation...
- Count Values within Cases...
- Shift Values...
- Recode into Same Variables...
- Recode into Different Variables...
- Automatic Recode...
- Create Dummy Variables
- Visual Binning...
- Optimal Binning...
- Prepare Data for Modeling >
- Rank Cases...
- Date and Time Wizard...
- Create Time Series...
- Replace Missing Values...
- Random Number Generators...
- Run Pending Transforms Ctrl+G



7 : Visible: 3 of 3 Variables

	id	var									
1	1										
2	2										
3	3										
4	4										
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File Edit View Data Transform Insert Format Analyze Graphs Utilities Extensions Window Help



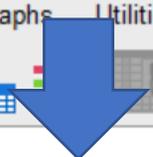
Output
Log

```
COMPUTE Difference=before - after.  
EXECUTE.
```



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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23														



Tada!

This new column represents the difference between the weights from the before and after columns!

Note the negative values: This is because I used:
before-after
I could get positive values if I used:
after-before
In truth, it makes no difference



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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19														
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22														
23														

We can explore whether our assumptions are met using the same approach as we did before



	id	before
1	1	187.2
2	2	194.2
3	3	231.7
4	4	200.5
5	5	201.7
6	6	235.0
7	7	208.7
8	8	172.4
9	9	184.6
10	10	189.6
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- Power Analysis >
- Meta Analysis >
- Reports >
- Descriptive Statistics >
- Bayesian Statistics >
- Tables >
- Compare Means >
- General Linear Model >
- Generalized Linear Models >
- Mixed Models >
- Correlate >
- Regression >
- Loglinear >
- Neural Networks >
- Classify >
- Dimension Reduction >
- Scale >
- Nonparametric Tests >
- Forecasting >
- Survival >
- Multiple Response >
- Missing Value Analysis...
- Multiple Imputation >
- Complex Samples >
- Simulation...
- Quality Control >
- Spatial and Temporal Modeling... >



- 123 Frequencies...
- Descriptives...
- Population Descript...
- Explore...
- Crosstabs...
- TURF Analysis
- Ratio...
- Proportion Confidence Intervals
- P-P Plots...
- Q-Q Plots...



We can explore whether our assumptions are met using the same approach as we did before



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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23														

Explore

Dependent List: Difference

Factor List:

Label Cases by:

Display: Both Statistics Plots

Buttons: Statistics..., Plots..., Options..., Bootstrap..., OK, Paste, Reset, Cancel, Help



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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17														
18														
19														
20														
21														
22														
23														

Explore

Dependent List:

- id
- before
- after

Explore: Statistics

- Descriptives
 - Confidence Interval for Mean: 95 %
- M-estimators
- Outliers
- Percentiles

Display: Both Statistics Plots

Buttons: Continue, Cancel, Help, OK, Paste, Reset, Cancel, Help



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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Explore: Plots

Boxplots

Factor levels together

Dependents together

None

Descriptive

Stem-and-leaf

Histogram

Normality plots with tests

Spread vs Level with Levene Test

None

Power estimation

Transformed Power: Natural log

Untransformed

Let's check our data together!



7 : Visible: 4 of 4 Variables

	id	before	after	Difference	var									
1	1	187.2	429.5	-242.30										
2	2	194.2	404.4	-210.20										
3	3	231.7	405.6	-173.90										
4	4	200.5	397.2	-196.70										
5	5	201.7	377.9	-176.20										
6	6	235.0	445.8	-210.80										
7	7	208.7	408.4	-199.70										
8	8	172.4	337.0	-164.60										
9	9	184.6	414.3	-229.70										
10	10	189.6	380.3	-190.70										
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23														

Time to run the analysis!



7 : Visible: 4 of 4 Variables

	id	before
1	1	187.2
2	2	194.2
3	3	231.7
4	4	200.5
5	5	201.7
6	6	235.0
7	7	208.7
8	8	172.4
9	9	184.6
10	10	189.6
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23		

Paired-Samples T Test

Paired Variables:

Pair	Variable1	Variable2
1		

Estimate effect sizes

Calculate standardizer using

- Standard deviation of the difference
- Corrected standard deviation of the difference
- Average of variances

Buttons: OK, Paste, Reset, Cancel, Help, Options..., Bootstrap...



7 : Visible: 4 of 4 Variables

	id	before
1	1	187.2
2	2	194.2
3	3	231.7
4	4	200.5
5	5	201.7
6	6	235.0
7	7	208.7
8	8	172.4
9	9	184.6
10	10	189.6
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19		
20		
21		
22		
23		

Paired-Samples T Test

id
before
after
Difference

Pair	Variable1	Variable2
1	[before]	[after]
2		

Estimate effect sizes

Calculate standardizer using

- Standard deviation of the difference
- Corrected standard deviation of the difference
- Average of variances

Options...
Bootstrap...

OK Paste Reset Cancel Help



7 : Visible: 4 of 4 Variables

	id	before
1	1	187.2
2	2	194.2
3	3	231.7
4	4	200.5
5	5	201.7
6	6	235.0
7	7	208.7
8	8	172.4
9	9	184.6
10	10	189.6
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21		
22		
23		

Paired-Samples T Test

id
before
after
Difference

Paired Variables:

Pair	Variable1	Variable2
1	[before]	[after]
2		

Estimate effect sizes

Calculate standardizer using

- Standard deviation of the difference
- Corrected standard deviation of the difference
- Average of variances

Options...
Bootstrap...

OK Paste Reset Cancel Help



Output

- T-Test
 - Title
 - Notes
 - Paired Samples Statistics
 - Paired Samples Correlations
 - Paired Samples Test
 - Paired Samples Effect Sizes

→ T-Test

Paired Samples Statistics

		Mean	N	Std. Deviation	Std. Error Mean
Pair 1	before	200.560	10	20.0282	6.3335
	after	400.040	10	30.0869	9.5143

Paired Samples Correlations

		N	Correlation	Significance	
				One-Sided p	Two-Sided p
Pair 1	before & after	10	.578	.040	.080

Paired Samples Test

		Paired Differences					Significance			
		Mean	Std. Deviation	Std. Error Mean	95% Confidence Interval of the Difference		t	df	One-Sided p	Two-Sided p
					Lower	Upper				
Pair 1	before - after	-199.4800	24.6929	7.8086	-217.1442	-181.8158	-25.546	9	<.001	<.001

Paired Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval		
				Lower	Upper	
				Pair 1	before - after	Cohen's d
		Hedges' correction	27.0200	-7.383	-10.776	-3.990

a. The denominator used in estimating the effect sizes.
Cohen's d uses the sample standard deviation of the mean difference.

The means for each group...

Worth noting as we should report this!



Output

- T-Test
 - Title
 - Notes
 - Paired Samples Statistics
 - Paired Samples Correlations
 - Paired Samples Test
 - Paired Samples Effect Sizes

T-Test

Paired Samples Statistics

		Mean	N	Std. Deviation	Std. Error Mean
Pair 1	before	200.560	10	20.0282	6.3335
	after	400.040	10	30.0869	9.5143

Paired Samples Correlations

		N	Correlation	Significance	
				One-Sided p	Two-Sided p
Pair 1	before & after	10	.578	.040	.080

Paired Samples Test

		Mean	Std. Deviation	Std. Error Mean	Paired Differences		t	df	Significance	
					95% Confidence Interval of the Difference				One-Sided p	Two-Sided p
					Lower	Upper				
Pair 1	before - after	-199.4800	24.6929	7.8086	-217.1442	-181.8158	-25.546	9	<.001	<.001

Paired Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval		
				Lower	Upper	
				Pair 1	before - after	Cohen's d
		Hedges' correction	27.0200	-7.383	-10.776	-3.990

a. The denominator used in estimating the effect sizes.
Cohen's d uses the sample standard deviation of the mean difference.

- First column: The pair of variables you are testing. The order the subtraction is indicated.
- Mean:** The average difference between the two treatments → also a good thing to report
- Standard deviation:** The standard deviation of the difference scores.
- Standard error mean:** The standard error (standard deviation divided by the square root of the sample size). Used in computing the test statistic



Output

- T-Test
 - Title
 - Notes
 - Paired Samples Statistics
 - Paired Samples Correlations
 - Paired Samples Test
 - Paired Samples Effect Sizes

→ T-Test

Paired Samples Statistics

		Mean	N	Std. Deviation	Std. Error Mean
Pair 1	before	200.560	10	20.0282	6.3335
	after	400.040	10	30.0869	9.5143

Paired Samples Correlations

		N	Correlation	Significance	
				One-Sided p	Two-Sided p
Pair 1	before & after	10	.578	.040	.080

Paired Samples Test

		Paired Differences					Significance			
		Mean	Std. Deviation	Std. Error Mean	95% Confidence Interval of the Difference		t	df	One-Sided p	Two-Sided p
					Lower	Upper				
Pair 1	before - after	-199.4800	24.6929	7.8086	-217.1442	-181.8158	-25.546	9	<.001	<.001

Paired Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Pair 1	before - after	Cohen's d	24.6929	-11.791	-4.366
		Hedges' correction	27.0200	-10.776	-3.990

a. The denominator used in estimating the effect sizes.
Cohen's d uses the sample standard deviation of the mean difference.

- t**: The test statistic
- df**: The degrees of freedom
- Sig. (2-tailed)**: The *p*-value corresponding to the given test statistic *t* with degrees of freedom *df*.



Output

- T-Test
 - Title
 - Notes
 - Paired Samples Statistics
 - Paired Samples Correlations
 - Paired Samples Test
 - Paired Samples Effect Sizes

→ T-Test

•Cohen's d: our measurement of effect size

Paired Samples Statistics

		Mean	N	Std. Deviation	Std. Error Mean
Pair 1	before	200.560	10	20.0282	6.3335
	after	400.040	10	30.0869	9.5143

Paired Samples Correlations

		N	Correlation	Significance	
				One-Sided p	Two-Sided p
Pair 1	before & after	10	.578	.040	.080

Paired Samples Test

		Paired Differences					Significance			
		Mean	Std. Deviation	Std. Error Mean	95% Confidence Interval of the Difference		t	df	One-Sided p	Two-Sided p
					Lower	Upper				
Pair 1	before - after	-199.4800	24.6929	7.8086	-217.1442	-181.8158	-25.546	9	<.001	<.001

Paired Samples Effect Sizes

		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Pair 1	before - after	Cohen's d	-8.078	-11.791	-4.366
		Hedges' correction	-7.383	-10.776	-3.990

a. The denominator used in estimating the effect sizes.
Cohen's d uses the sample standard deviation of the mean difference.

We have a significant result! So what do we report?

- The average weight of the 10 rats in our sample before the food court opened up was 200.56g (SD=20.03), and increased to 400.04g (SD=30.09) after the food court opened. A paired samples T Test revealed this difference to be significant, $t(9) = 25.55$, $p < 0.001$, $d = 8.078$.
- Notice I removed the negative from cohen's d. This is fine

If normality isn't met

- Just like with our 2 samples T Test, there is a non-parametric equivalent for when our data do not meet the required assumption of normality
- Lets open the data 'rats3.xlsx'



Visible: 3 of 3 Variables

	id	Before	After	var											
1	1	200.1	200.5												
2	2	201.9	200.8												
3	3	192.7	198.7												
4	4	216.1	227.7												
5	5	170.0	271.8												
6	6	196.9	265.5												
7	7	172.2	259.6												
8	8	185.5	288.5												
9	9	160.3	249.0												
10	10	193.7	189.8												
11															
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Let's say you run a similar experiment in a different town with a different food court (rats were different to start off with, and the choice of garbage they can eat after the mall opens up is different)

Let's run the same diagnostics to check for normality as we have been in the past...

- **the difference between the two groups** are not normally distributed
- This means we cannot use a paired samples T Test!
- We do have a non-parametric equivalent though!

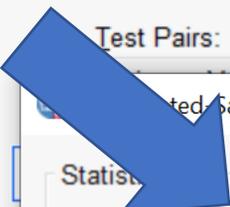
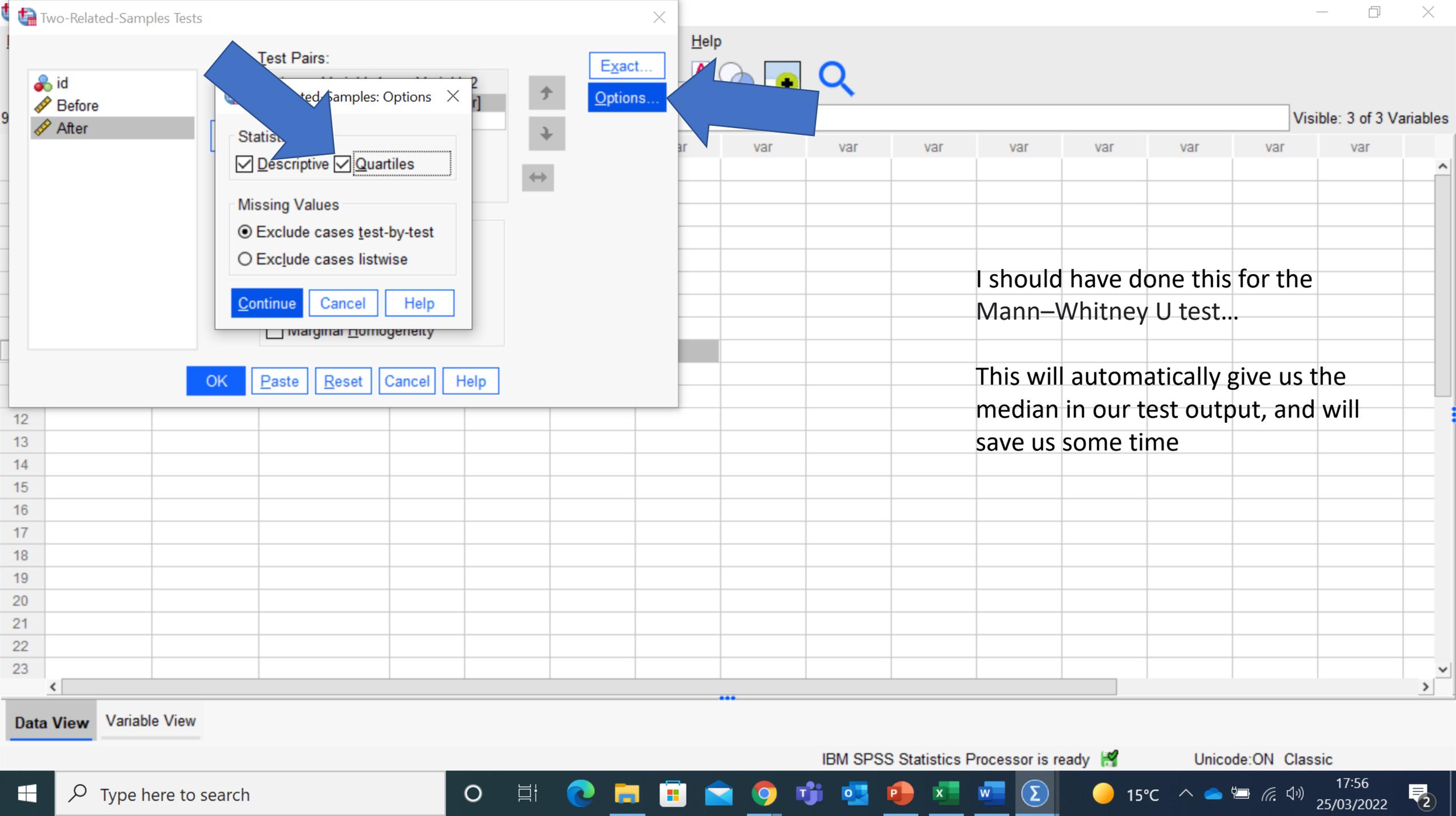
The **Wilcoxon signed rank test for paired samples**



Visible: 3 of 3 Variables

	id	Before	After	var											
1	1	200.1	200.5												
2	2	201.9	200.8												
3	3	192.7	198.7												
4	4	216.1	227.7												
5	5	170.0	271.8												
6	6	196.9	265.5												
7	7	172.2	259.6												
8	8	185.5	288.5												
9	9	160.3	249.0												
10	10	193.7	189.8												
11															
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22															
23															

Data View Variable View



I should have done this for the Mann–Whitney U test...

This will automatically give us the median in our test output, and will save us some time

Output

- NPAr Tests
 - Title
 - Notes
 - Descriptive Statistics
 - Wilcoxon Signed R
 - Title
 - Ranks
 - Test Statistics

NPAr Tests

Descriptive Statistics

	N	Mean	Deviation	Minimum	Maximum	25th	Percentiles 50th (Median)	75th
Before	10	188.94	17.0012	160.3	216.1	171.650	193.200	200.550
After	10	235.190	36.0885	189.8	288.5	200.050	238.350	267.075

Wilcoxon Signed Ranks

Ranks

	N	Mean Rank	Sum of Ranks
After - Before			
Negative Ranks	2 ^a	2.50	5.00
Positive Ranks	8 ^b	6.25	50.00
Ties	0 ^c		
Total	10		

- a. After < Before
- b. After > Before
- c. After = Before

Test Statistics^a

	After - Before
Z	-2.293 ^b
Asymp. Sig. (2-tailed)	.022

a. Wilcoxon Signed Ranks Test

The **Ranks** table provides data on the comparison of participants' Before and After weight.

Notice legend at the bottom of table:
 2 rats had lower weights after the treatment vs before
 8 rats had higher weights after the treatment vs before
 0 rats maintained the same weight

Output

- ↳ NPar Tests
 - ↳ Title
 - ↳ Notes
 - ↳ Descriptive Statistics
 - ↳ Wilcoxon Signed R
 - ↳ Title
 - ↳ Ranks
 - ↳ Test Statistics

↳ NPar Tests

Descriptive Statistics

	N	Mean	Std. Deviation	Minimum	Maximum	25th	Percentiles 50th (Median)	75th
Before	10	188.940	17.0012	160.3	216.1	171.650	193.200	200.550
After	10	235.190	36.0885	189.8	288.5	200.050	238.350	267.075

Wilcoxon Signed Ranks Test

Ranks

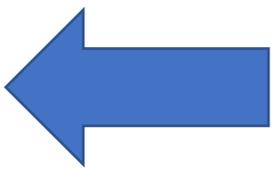
		N	Mean Rank	Sum of Ranks
After - Before	Negative Ranks	2 ^a	2.50	5.00
	Positive Ranks	8 ^b	6.25	50.00
	Ties	0 ^c		
	Total	10		

- a. After < Before
- b. After > Before
- c. After = Before

Test Statistics^a

	After - Before
Z	-2.293 ^b
Asymp. Sig. (2-tailed)	.022

a. Wilcoxon Signed Ranks Test



It seems we have a significant p value, suggesting to us there was a significant change in weight between our two sampling periods!

What do we report?

- Our data indicate that rats in the neighbourhood were heavier after the vegan restaurant opened (MDN = 238.35) compared to baseline measurements beforehand (MDN = 193.20). A Wilcoxon Signed-ranks test indicated that there was a significant difference in rat weight across the two sample periods, $z = -2.293$, $p = 0.022$.

What if we don't have a continuous response variable?

- In all four of the statistical analyses we've looked at so far, we have been looking at the effects of a dichotomous independent variable (2 classes) on a continuous dependant variable
- Effect of sex (2 levels) on body weight (continuous)
- Effect of sex (2 levels) on number of aggressive interactions over a study period (continuous)
- Effect of sampling period (**before** vs **after** restaurant opening) on body weight (continuous)

What if we don't have a continuous response variable?

- What if we are interested in how a categorical variable effects another categorical variable?

Example

- You are running a natural experiment:
- You have 2 age classes of arboreal monkeys (Capuchin monkeys in Costa Rica - immature and mature)
- You are interested whether immature and mature individuals bias their foraging to different parts of a tree crown
 - Maybe smaller immatures want to avoid being on the edge of the tree crown, as birds of prey are adept at catching small, inexperienced youngsters
 - Alternatively, it may be better for smaller animals to be at the edge of the tree crown, because there is less feeding competition at the edge from larger monkeys that don't allow them to eat enough!

- As a wildlife biologist, you are extremely interested in knowing how age affects where these monkeys are foraging.
- You can easily identify whether a monkey is an adult versus as immature based on the characteristics of their genitalia
- You also come up with a rudimentary measure of whether the monkey is at the 'centre' of a tree crown, or at the 'edge'

- Let's open the data 'Monkey_tree1.xlsx'



Visible: 3 of 3 Variables

	V1	Age	Position	var												
1	51	Imm	Edge													
2	52	Imm	Edge													
3	53	Imm	Edge													
4	54	Imm	Center													
5	55	Imm	Edge													
6	56	Imm	Center													
7	57	Imm	Edge													
8	58	Imm	Center													
9	59	Imm	Edge													
10	60	Imm	Center													
11	61	Imm	Center													
12	62	Imm	Edge													
13	63	Imm	Edge													
14	64	Imm	Edge													
15	65	Imm	Center													
16	66	Imm	Edge													
17	67	Imm	Center													
18	68	Imm	Edge													
19	69	Imm	Edge													
20	70	Imm	Center													
21	71	Imm	Edge													
22	72	Imm	Edge													

Data View Variable View

Assumptions

- **Assumption 1: Both variables are categorical**
 - They sure are in this case!
- **Assumption 2: All observations are independent**
 - The value of one measurement does not depend on the other
- **Assumption 3: Expected value of cells should be 5 or greater in at least 80% of cells.**
 - We will check this on our output



Visible: 3 of 3 Variables

	V1	Age	Position	var												
1	51	Imm	Edge													
2	52	Imm	Edge													
3	53	Imm	Edge													
4	54	Imm	Center													
5	55	Imm	Edge													
6	56	Imm	Center													
7	57	Imm	Edge													
8	58	Imm	Center													
9	59	Imm	Edge													
10	60	Imm	Center													
11	61	Imm	Center													
12	62	Imm	Edge													
13	63	Imm	Edge													
14	64	Imm	Edge													
15	65	Imm	Center													
16	66	Imm	Edge													
17	67	Imm	Center													
18	68	Imm	Edge													
19	69	Imm	Edge													
20	70	Imm	Center													
21	71	Imm	Edge													
22	72	Imm	Edge													

Data View Variable View



	V1	Age	Position	var	var
1	51	Imm	Edge		
2	52	Imm	Edge		
3	53	Imm	Edge		
4	54	Imm	Center		
5	55	Imm	Edge		
6	56	Imm	Center		
7	57	Imm	Edge		
8	58	Imm	Center		
9	59	Imm	Edge		
10	60	Imm	Center		
11	61	Imm	Center		
12	62	Imm	Edge		
13	63	Imm	Edge		
14	64	Imm	Edge		
15	65	Imm	Center		
16	66	Imm	Edge		
17	67	Imm	Center		
18	68	Imm	Edge		
19	69	Imm	Edge		
20	70	Imm	Center		
21	71	Imm	Edge		
22	72	Imm	Edge		

Visible: 3 of 3 Variables

Crosstabs

Row(s):
Age

Column(s):
Position

Layer 1 of 1

Previous Next

Display clustered bar charts

Suppress tables

OK Paste Reset Cancel Help

Exact...
Statistics...
Cells...
Format...
Style...
Bootstrap...



Visible: 3 of 3 Variables

	V1	Age	Position	var	var
1	51	Imm	Edge		
2	52	Imm	Edge		
3	53	Imm	Edge		
4	54	Imm	Center		
5	55	Imm	Edge		
6	56	Imm	Center		
7	57	Imm	Edge		
8	58	Imm	Center		
9	59	Imm	Edge		
10	60	Imm	Center		
11	61	Imm	Center		
12	62	Imm	Edge		
13	63	Imm	Edge		
14	64	Imm	Edge		
15	65	Imm	Center		
16	66	Imm	Edge		
17	67	Imm	Center		
18	68	Imm	Edge		
19	69	Imm	Edge		
20	70	Imm	Center		
21	71	Imm	Edge		
22	72	Imm	Edge		

Crosstabs

V1

Crosstabs: Statistics

- Chi-square
- Correlations
- Nominal**
 - Contingency coefficient
 - Phi and Cramer's V
 - Lambda
 - Uncertainty coefficient
- Ordinal**
 - Gamma
 - Somers' d
 - Kendall's tau-b
 - Kendall's tau-c
- Nominal by Interval**
 - Eta
- Kappa
- Risk
- McNemar
- Cochran's and Mantel-Haenszel statistics

Test common odds ratio equals: 1

Display clustered bar charts

Suppress table

Continue Cancel Help

OK Paste Reset Cancel Help

Exact...

Statistics...

Cells...

Format...

Style...

Bootstrap...



	V1	Age	Position	var	var
1	51	Imm	Edge		
2	52	Imm	Edge		
3	53	Imm	Edge		
4	54	Imm	Center		
5	55	Imm	Edge		
6	56	Imm	Center		
7	57	Imm	Edge		
8	58	Imm	Center		
9	59	Imm	Edge		
10	60	Imm	Center		
11	61	Imm	Center		
12	62	Imm	Edge		
13	63	Imm	Edge		
14	64	Imm	Edge		
15	65	Imm	Center		
16	66	Imm	Edge		
17	67	Imm	Center		
18	68	Imm	Edge		
19	69	Imm	Edge		
20	70	Imm	Center		
21	71	Imm	Edge		
22	72	Imm	Edge		

Crosstabs: Cell Display

Counts

- Observed
- Expected
- Hide small counts
Less than 5

z-test

- Compare column proportions
- Adjust p-values (Bonferroni method)

Percentages

- Row
- Column
- Total

Residuals

- Unstandardized
- Standardized
- Adjusted standardized

Noninteger Weights

- Round cell counts
- Round case weights
- Truncate cell counts
- Truncate case weights
- No adjustments

Continue Help

Click Cells (not easily seen in this screen-shot)



	V1	Age	Position	var	var
1	51	Imm	Edge		
2	52	Imm	Edge		
3	53	Imm	Edge		
4	54	Imm	Center		
5	55	Imm	Edge		
6	56	Imm	Center		
7	57	Imm	Edge		
8	58	Imm	Center		
9	59	Imm	Edge		
10	60	Imm	Center		
11	61	Imm	Center		
12	62	Imm	Edge		
13	63	Imm	Edge		
14	64	Imm	Edge		
15	65	Imm	Center		
16	66	Imm	Edge		
17	67	Imm	Center		
18	68	Imm	Edge		
19	69	Imm	Edge		
20	70	Imm	Center		
21	71	Imm	Edge		
22	72	Imm	Edge		

Visible: 3 of 3 Variables

Crosstabs

Row(s): Age

Column(s): Position

Layer 1 of 1

Previous Next

Display clustered bar charts

Suppress tables

OK Paste Reset Cancel Help

Exact... Statistics... Cells... Format... Style... Bootstrap...

Output

- GGraph
 - Title
 - Notes
 - Active Dataset
 - Graph
- Crosstabs
 - Title
 - Notes
 - Case Processing Summary
 - Age * Position Crosstabulation
 - Chi-Square Tests
 - Symmetric Measures

Case Processing Summary

	Valid		Cases Missing		Total	
	N	Percent	N	Percent	N	Percent
Age * Position	100	100.0%	0	0.0%	100	100.0%

Age * Position Crosstabulation

		Position		
		Center	Edge	Total
Age	Adult	Count: 6	Count: 44	Count: 50
		Expected Count: 13.5	Expected Count: 36.5	Expected Count: 50.0
	Imm	Count: 21	Count: 29	Count: 50
		Expected Count: 13.5	Expected Count: 36.5	Expected Count: 50.0
Total		Count: 27	Count: 73	Count: 100
		Expected Count: 27.0	Expected Count: 73.0	Expected Count: 100.0

Chi-Square Tests

	Value	df	Asymptotic Significance (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)
Pearson Chi-Square	11.416 ^a	1	<.001		
Continuity Correction ^b	9.944	1	.002		
Likelihood Ratio	11.930	1	<.001		
Fisher's Exact Test				.001	<.001
N of Valid Cases	100				

a. 0 cells (0.0%) have expected count less than 5. The minimum expected count is 13.50.

Crosstabulation table:
Shows us the actual count for each category...
So there were 6 instances where we found Adults at the 'centre' of the tree crown

There were 44 instances where we saw an adult at the edge!

We also have an **expected value**, assuming no association between age and foraging position

: Based on the numbers of adults we observed, and the number of 'edge' observations we observed, we would expect 36.5 instances of adults at the edge!



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Case Processing Summary

	Valid		Missing		Total	
	N	Percent	N	Percent	N	Percent
Age * Position	100	100.0%	0	0.0%	100	100.0%

Age * Position Crosstabulation

		Position			
		Center	Edge	Total	
Age	Adult	Count	6	44	50
		Expected Count	13.5	36.5	50.0
Imm	Imm	Count	21	29	50
		Expected Count	13.5	36.5	50.0
Total		Count	27	73	100
		Expected Count	27.0	73.0	100.0



Are the differences between expected and actual values different enough to say something is going on?

Chi-Square Tests

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Fisher's Exact Test				.001	<.001
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Age * Position Crosstabulation

		Position		Total	
		Center	Edge		
Age	Adult	Count	6	44	50
		Expected Count	13.5	36.5	50.0
	Imm	Count	21	29	50
		Expected Count	13.5	36.5	50.0
Total		Count	27	73	100
		Expected Count	27.0	73.0	100.0

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Fisher's Exact Test				.001	<.001
N of Valid Cases	100				

a. 0 cells (0.0%) have expected count less than 5. The minimum expected count is 13.50.

b. Computed only for a 2x2 table



Symmetric Measures

		Value	Approximate Significance
Nominal by Nominal	Phi	-.338	<.001
	Cramer's V	.338	<.001

Note: Chisquared needs enough data so that the 'expected' value is above 5 in at least 80% of cells

So here, we need this % of values with expected counts less than 5 to be less than 20%

We are A-okay here!



- Output
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Age * Position Crosstabulation

		Position			
		Center	Edge	Total	
Age	Adult	Count	6	44	50
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Imm	Count	21	29	50	
		Expected Count	13.5	36.5	50.0
Total	Count	27	73	100	
		Expected Count	27.0	73.0	100.0

Our chi squared value is 11.416

Our degrees of freedom is 1

Chi-Square Tests

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		Center	Edge	Total	
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		Expected Count	13.5	36.5	50.0
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Total	Count	27	73	100	
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Symmetric Measures

		Value	Approximate Significance
Nominal by Nominal	Phi	-.338	<.001
	Cramer's V	.338	<.001

Our p value is well below 0.05

We can accept our alternate hypothesis that there IS a significant relationship between age category and 'foraging position' within the tree





Output

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Fisher's Exact Test				.001	<.001
N of Valid Cases	100				

a. 0 cells (0.0%) have expected count less than 5. The minimum expected count is 13.50.
 b. Computed only for a 2x2 table

Symmetric Measures

	Value	Asymptotic Significance
Nominal by Nominal	Phi	-.338
	Cramer's V	.338
N of Valid Cases		100



For effect size, we can look at our value for 'Phi'

As was the case with cohen's d, there are some general rules of thumb available for this. These can vary from textbook to textbook. My current favourite text uses:

0.1 is considered a small effect, 0.3 a medium effect and 0.5 a large effect.

- χ^2 (degrees of freedom, N = sample size) = chi-square statistic value, p = p value.
- Example
- A chi-square test of independence was performed to examine the relation between age group and foraging position within tree crowns. The relation between these variables was significant, $\chi^2 (1, N = 100) = 11.416, p < .001$.

Let's make a figure!



- Chart Builder...
- Graphboard Template Chooser...
- Relationship Map...
- Weibull Plot...
- Compare Subgroups
- Regression Variable Plots



24 :

	V1	Age	Position	var
1	51	Imm	Edge	
2	52	Imm	Edge	
3	53	Imm	Edge	
4	54	Imm	Center	
5	55	Imm	Edge	
6	56	Imm	Center	
7	57	Imm	Edge	
8	58	Imm	Center	
9	59	Imm	Edge	
10	60	Imm	Center	
11	61	Imm	Center	
12	62	Imm	Edge	
13	63	Imm	Edge	
14	64	Imm	Edge	
15	65	Imm	Center	
16	66	Imm	Edge	
17	67	Imm	Center	
18	68	Imm	Edge	
19	69	Imm	Edge	
20	70	Imm	Center	
21	71	Imm	Edge	
22	72	Imm	Edge	
23	73	Imm	Edge	
24	74	Imm	Edge	



Legacy Dialogs >

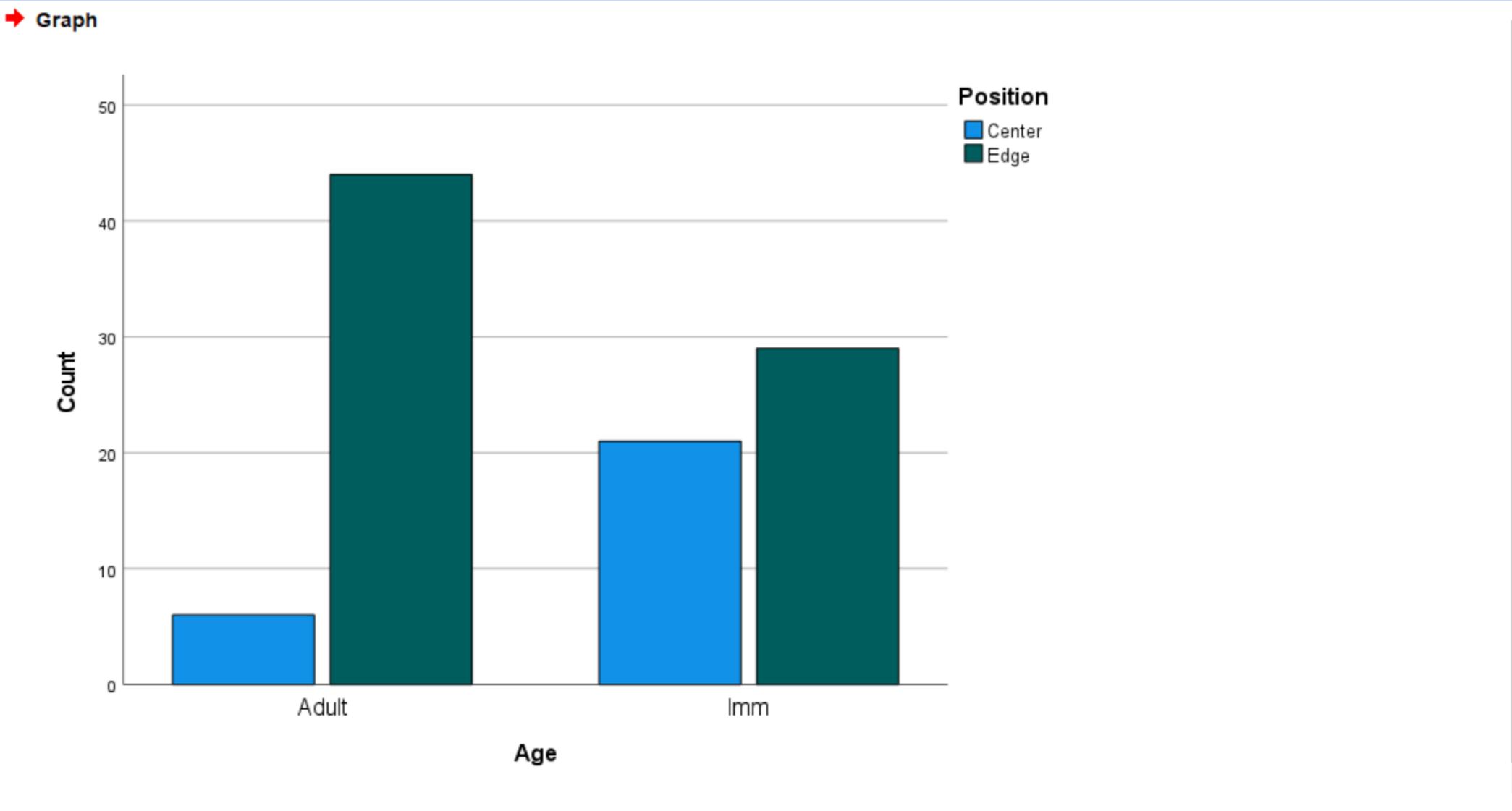
- Bar...
- 3-D Bar...
- Line...
- Area...
- Pie...
- High-Low...
- Boxplot...
- Error Bar...
- Population Pyramid...
- Scatter/Dot...
- Histogram...

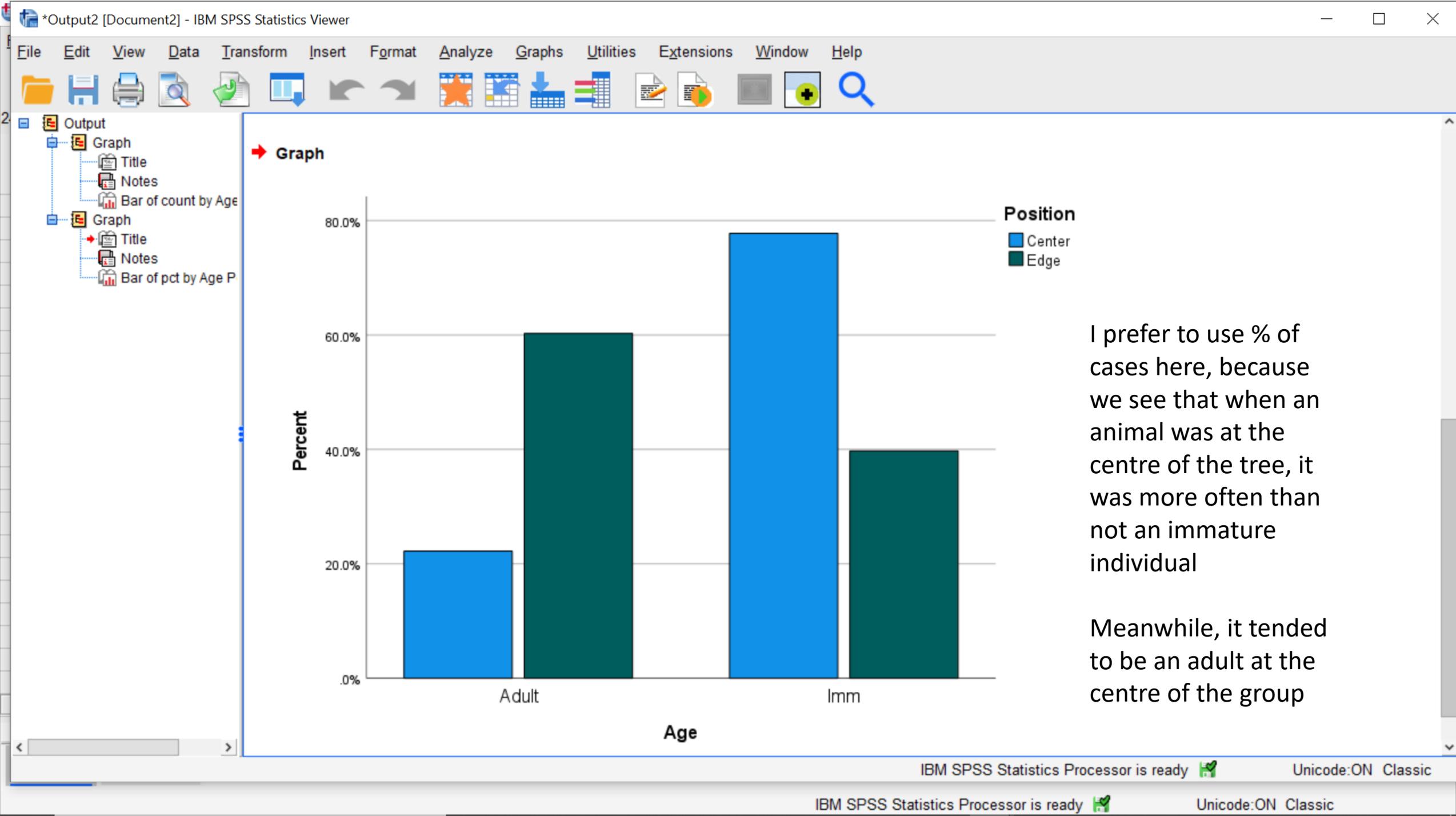


Visible: 3 of 3 Variables



- Output
 - Graph
 - Title
 - Notes
 - Bar of count by Age





I prefer to use % of cases here, because we see that when an animal was at the centre of the tree, it was more often than not an immature individual

Meanwhile, it tended to be an adult at the centre of the group



	Name	Type	Width
1	V1	Numeric	3
2	Age	String	5
3	Position	String	6
4			
5			
6			
7			
8			
9			
10			
11			
12			
13			
14			
15			
16			
17			
18			
19			
20			
21			
22			
23			
24			
25			

Chart Builder X

Chart preview uses example data

Variables:

V1

Age

Position

No categories (scale variable)

Drag a Gallery chart here to use it as your starting point

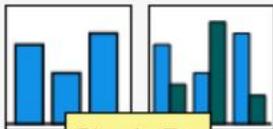
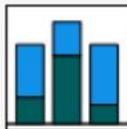
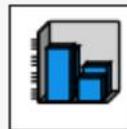
OR

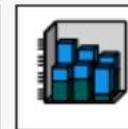
Click on the Basic Elements tab to build a chart element by element

Gallery Basic Elements Groups/Point ID Titles/Footnotes

Choose from:

- Favorites
- Bar
- Line
- Area
- Pie/Polar
- Scatter/Dot
- Histogram
- High-Low
- Boxplot
- Dual Axes



Element Properties Chart Appearance Options

Edit Properties of:



	Name	Type	Width
1	V1	Numeric	3
2	Age	String	5
3	Position	String	6
4			
5			
6			
7			
8			
9			
10			
11			
12			
13			
14			
15			
16			
17			
18			
19			
20			
21			
22			
23			
24			
25			

Chart Builder X

Variables: Chart preview uses example data

V1

Age

Position

No categories (scale variable)

Simple Bar Count

Y-Axis?

X-Axis?

Filter?

Gallery Basic Elements Groups/Point ID Titles/Footnotes

Choose from:

- Favorites
- Bar
- Line
- Area
- Pie/Polar
- Scatter/Dot
- Histogram
- High-Low
- Boxplot
- Dual Axes

Simple Bar Stacked 3-D Bar

Element Properties Chart Appearance Options

Edit Properties of:

Bar1

X-Axis1 (Bar1)

Y-Axis1 (Bar1)

Statistics

Variable:

Statistic:

Count

Set Parameters...

Display error bars

Error Bars Represent

Confidence intervals

Level (%): 95

Standard error

Multiplier: 2

Standard deviation

Multiplier: 2

Bar Style:

Bar

- You sit down with Dr. Lynn Besenyei, who teaches you how to bait and use traps for small mammals
- For your masters project, you spend your time traveling to your field sites, capturing mice, weighing them, marking them, and releasing them

