

Analysis of Quantitative data One-Way + Two-Way ANOVA

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Comparison between more than 2 groups One factor = One predictor One-Way ANOVA

Signal-to-noise ratio

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Signal – Difference between the means
Noise – Variability in the groups
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= F ratio

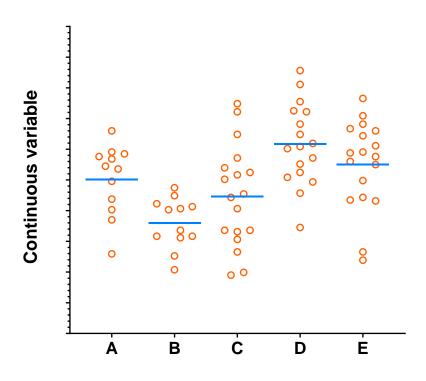
One-Way Analysis of variance

Step 1: Omnibus test

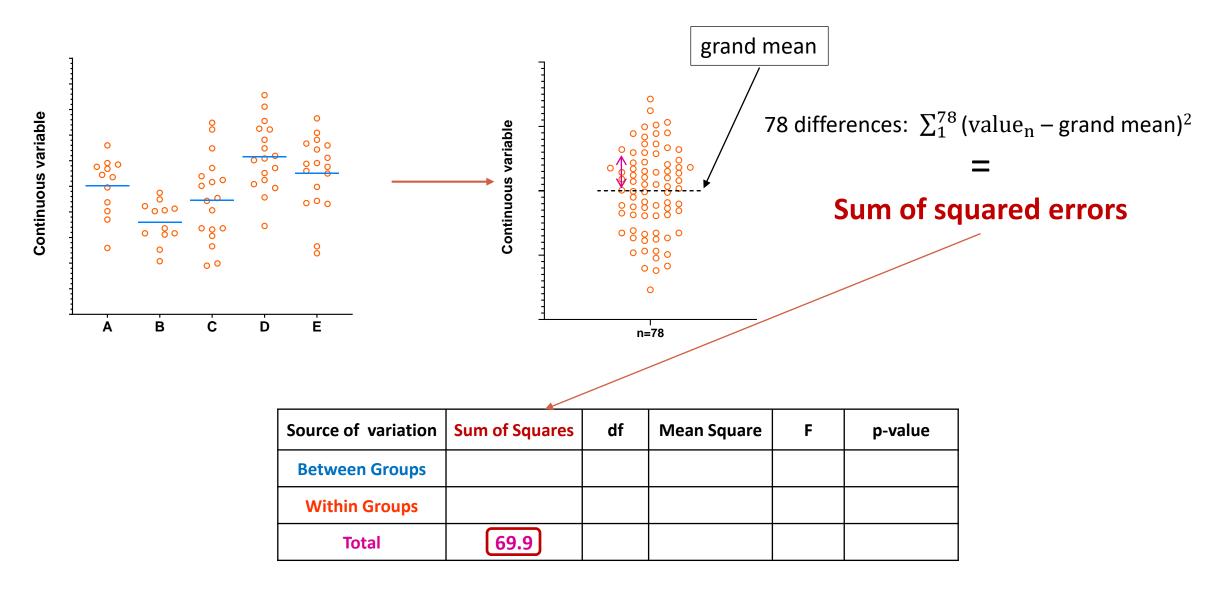
 It tells us if there is a difference between the means but not which means are significantly different from which other ones.

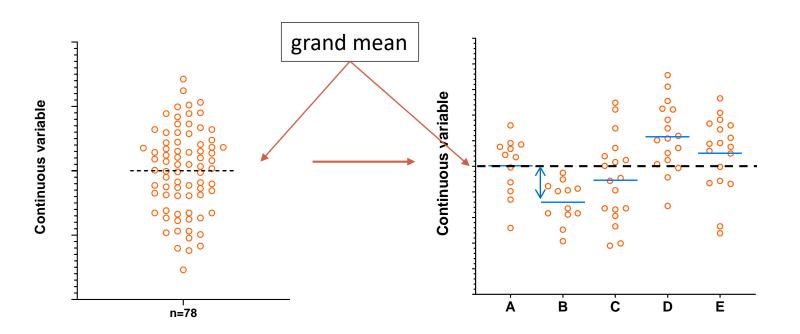
Step 2: Post-hoc tests

They tell us if there are differences between the means pairwise.



Source of variation	Sum of Squares	df	Mean Square	F	p-value
Between Groups	18.1	4	4.5	6.32	0.0002
Within Groups	51.8	73	0.71		
Total	69.9				



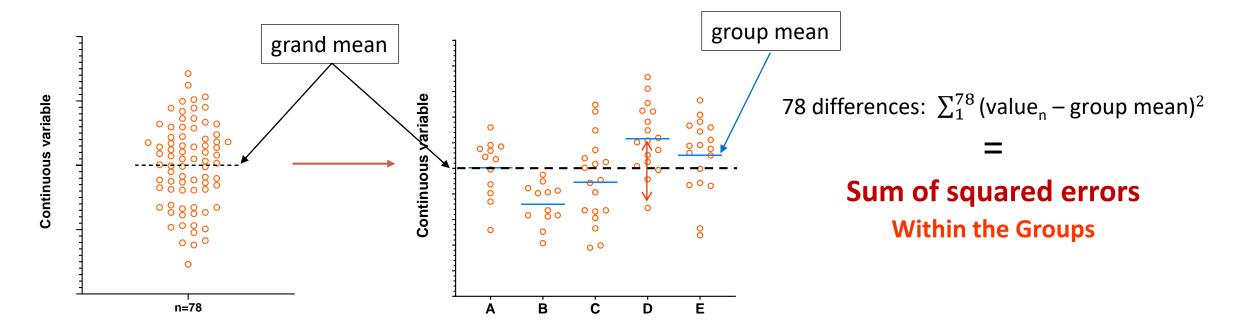


5 differences: $\sum_{1}^{5} (\text{mean}_{n} - \text{grand mean})^{2}$

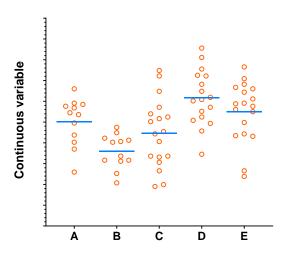
Sum of squared errors

Between the groups

Source of variation	Sum of Squares	df	Mean Square	F	p-value
Between Groups	18.1				
Within Groups					
Total	69.9				



Source of variation	Sum of Squares	df	Mean Squares	F	p-value
Between Groups	18.1				
Within Groups	51.8				
Total	69.9				



Signal Noise

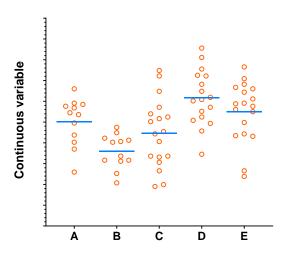
Source of variation	Sum of Squares	df	Mean Squares	F ratio	p-value
Between Groups	18.1	k-1			
Within Groups	51.8	n-k			
Total	69.9				

df: degree of freedom with df = n-1

n = number of values, k=number of groups

Between groups: df = 4 (k-1)

Within groups: $df = 73 (n-k = n_1-1 + ... + n_5-1)$

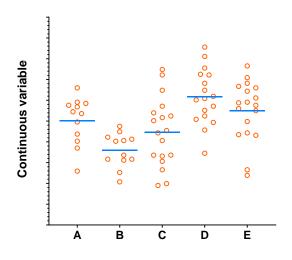


Signal Noise

Source of variation	Sum of Squares	df	Mean Squares	F ratio	p-value
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Total	69.9				

df: degree of freedom with df = n-1

Mean squares = Sum of Squares / n-1 = Variance!



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Mean squares = Sum of Squares / n-1 = Variance

Fratio =
$$\frac{\text{Variance between the groups}}{\text{Variance within the groups (individual variability)}} = \frac{4.5}{0.71} = 6.34$$

Comparison of more than 2 means

- Running multiple tests on the same data increases the **familywise error rate**.
- What is the familywise error rate?
 - The error rate across tests conducted on the same experimental data.
- One of the basic rules ('laws') of probability:
 - The Multiplicative Rule: The probability of the joint occurrence of 2 or more independent events is the product of the individual probabilities.

$$P(A,B) = P(A) \times P(B)$$

For example:

 $P(2 \text{ Heads}) = P(\text{head}) \times P(\text{head}) = 0.5 \times 0.5 = 0.25$

Familywise error rate

- **Example**: All pairwise comparisons between 3 groups A, B and C:
 - A-B, A-C and B-C
- Probability of making the Type I Error: 5%
 - The probability of <u>not making the Type I Error</u> is 95% (=1 0.05)
- Multiplicative Rule:
 - Overall probability of no Type I errors is: 0.95 * 0.95 * 0.95 = 0.857
- So the probability of making at least one Type I Error is 1-0.857 = 0.143 or **14.3**%
 - The probability has increased from 5% to 14.3%
- Comparisons between 5 groups instead of 3, the familywise error rate is 40% (=1-(0.95)ⁿ)

Familywise error rate

- Solution to the increase of familywise error rate: correction for multiple comparisons
 - Post-hoc tests
- Many different ways to correct for multiple comparisons:
 - Different statisticians have designed corrections addressing different issues
 - e.g. unbalanced design, heterogeneity of variance, liberal vs conservative
- However, they all have one thing in common:
 - the more tests, the higher the familywise error rate: the more stringent the correction
- Tukey, Bonferroni, Sidak, Benjamini-Hochberg ...
 - Two ways to address the multiple testing problem
 - Familywise Error Rate (FWER) vs. False Discovery Rate (FDR)

Multiple testing problem

- **FWER**: **Bonferroni**: $\alpha_{adiust} = 0.05/n$ comparisons e.g. 3 comparisons: 0.05/3=0.016
 - Problem: very conservative leading to <u>loss of power</u> (lots of false negative)
 - 10 comparisons: threshold for significance: 0.05/10: 0.005
 - Pairwise comparisons across 20.000 genes ☺
- <u>FDR</u>: Benjamini-Hochberg: the procedure controls the expected proportion of "discoveries" (significant tests) that are false (false positive).
 - Less stringent control of Type I Error than FWER procedures which control the probability of <u>at least</u> one Type I Error
 - More power at the cost of increased numbers of Type I Errors.

Difference between FWER and FDR:

- a p-value of 0.05 implies that 5% of all tests will result in false positives.
- a FDR adjusted p-value (or q-value) of 0.05 implies that 5% of significant tests will result in false positives.

One-Way Analysis of variance

Step 1: Omnibus test

• It tells us if there is (or not) a difference between the means but not which means are significantly different from which other ones.

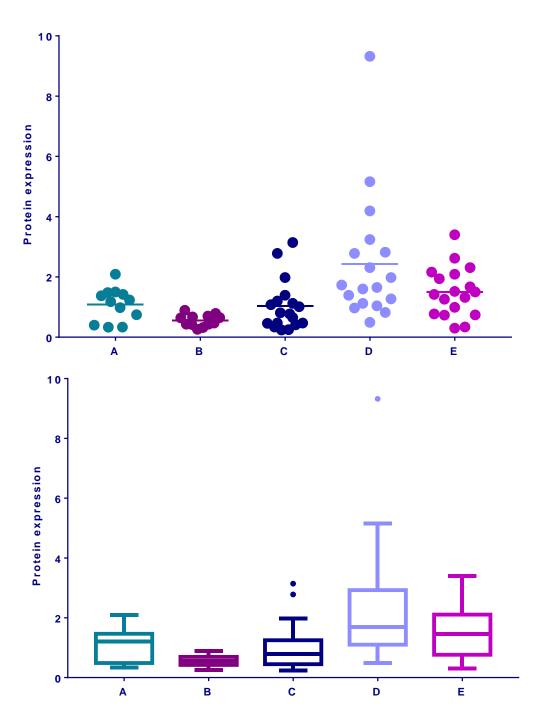
Step 2: Post-hoc tests

- They tell us if there are (or not) differences between the means pairwise.
- A correction for multiple comparisons will be applied on the p-values.
- These post hoc tests should only be used when the ANOVA finds a significant effect.

Exercise: One-way ANOVA protein expression.xlsx

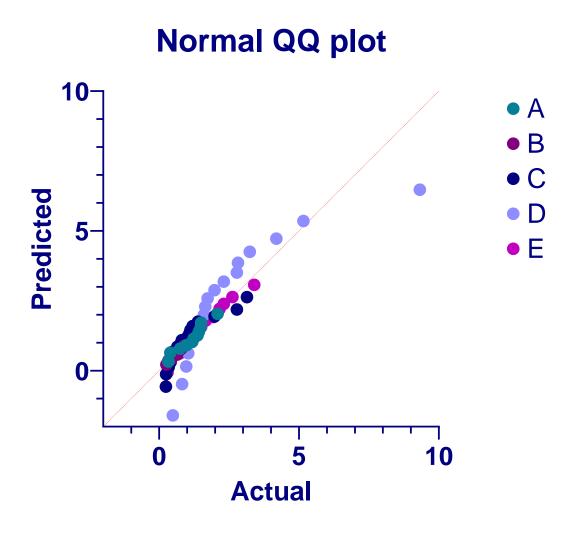
 Question: is there a difference in protein expression between the 5 cell lines?

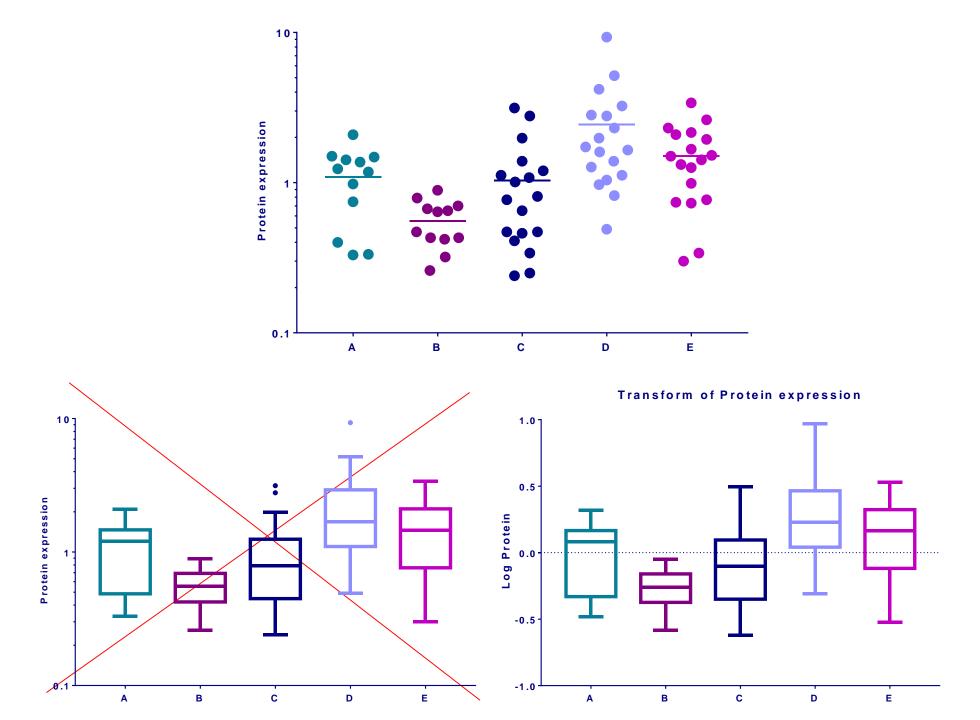
- 1 Plot the data
- 2 Check the assumptions for parametric test



Parametric tests assumptions

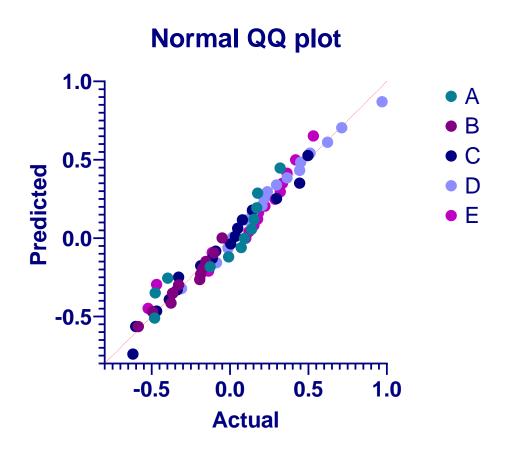
4						
1	Test for normal distribution					
2	Anderson-Darling test				_	
3	A2*	0.3797	0.3141	1.166	1.439	0.2011
4	P value	0.3446	0.5029	0.0035	0.0007	0.8590
5	Passed normality test (alpha=0.05)?	Yes	Yes	No	No	Yes
6	P value summary	ns	ns	**	***	ns
7						
8	D'Agostino & Pearson test					
9	K2	0.1236	0.7508	9.375	22.59	1.280
10	P value	0.9401	0.6870	0.0092	<0.0001	0.5274
11	Passed normality test (alpha=0.05)?	Yes	Yes	No	No	Yes
12	P value summary	ns	ns	**	****	ns
13						
14	Shapiro-Wilk test					
15	W	0.9295	0.9535	0.8197	0.7531	0.9671
16	P value	0.3752	0.6888	0.0029	0.0004	0.7411
17	Passed normality test (alpha=0.05)?	Yes	Yes	No	No	Yes
18	P value summary	ns	ns	**	***	ns
19						
20	Kolmogorov-Smirnov test					
21	KS distance	0.1485	0.1704	0.1980	0.2058	0.1035
22	P value	>0.1000	>0.1000	0.0603	0.0424	>0.1000
23	Passed normality test (alpha=0.05)?	Yes	Yes	Yes	No	Yes
24	P value summary	ns	ns	ns	*	ns
25						
26	Number of values	12	12	18	18	18



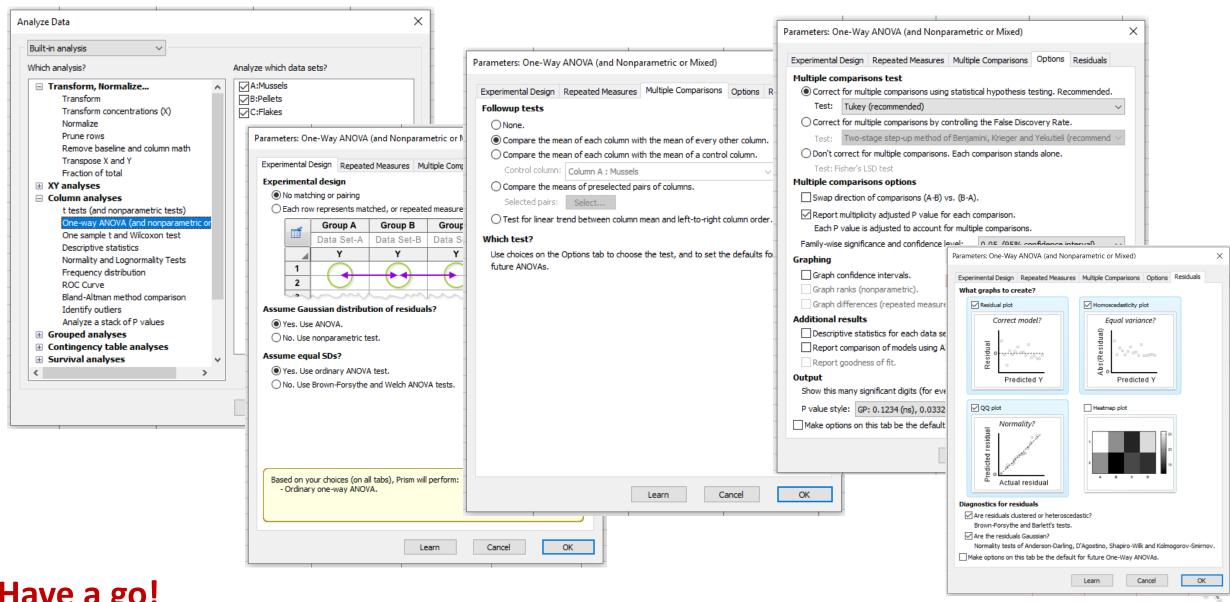


Parametric tests assumptions

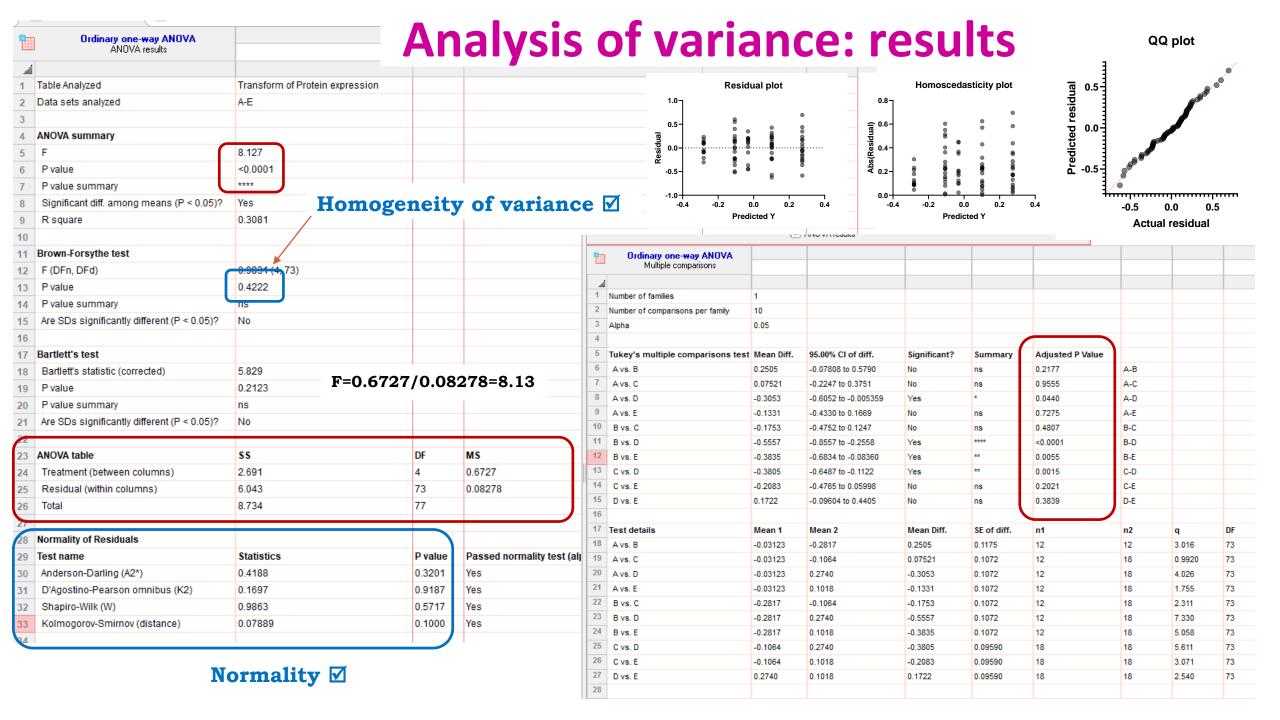
1	Test for normal distribution					
2	Anderson-Darling test					
3	A2*	0.7849	0.3412	0.2086	0.1524	0.4727
4	P value	0.0295	0.4303	0.8386	0.9495	0.2138
5	Passed normality test (alpha=0.05)?	No	Yes	Yes	Yes	Yes
6	P value summary	*	ns	ns	ns	ns
7						
8	D'Agostino & Pearson test					
9	K2	2.037	0.6827	0.5884	0.8869	2.902
10	P value	0.3611	0.7108	0.7451	0.6418	0.2344
11	Passed normality test (alpha=0.05)?	Yes	Yes	Yes	Yes	Yes
12	P value summary	ns	ns	ns	ns	ns
13						
14	Shapiro-Wilk test					
15	W	0.8553	0.9458	0.9657	0.9868	0.9313
16	P value	0.0427	0.5773	0.7142	0.9935	0.2050
17	Passed normality test (alpha=0.05)?	No	Yes	Yes	Yes	Yes
18	P value summary	*	ns	ns	ns	ns
19						
20	Kolmogorov-Smirnov test					
21	KS distance	0.2278	0.2049	0.1373	0.1016	0.1646
22	P value	0.0857	>0.1000	>0.1000	>0.1000	>0.1000
23	Passed normality test (alpha=0.05)?	Yes	Yes	Yes	Yes	Yes
24	P value summary	ns	ns	ns	ns	ns
25						
26	Number of values	12	12	18	18	18



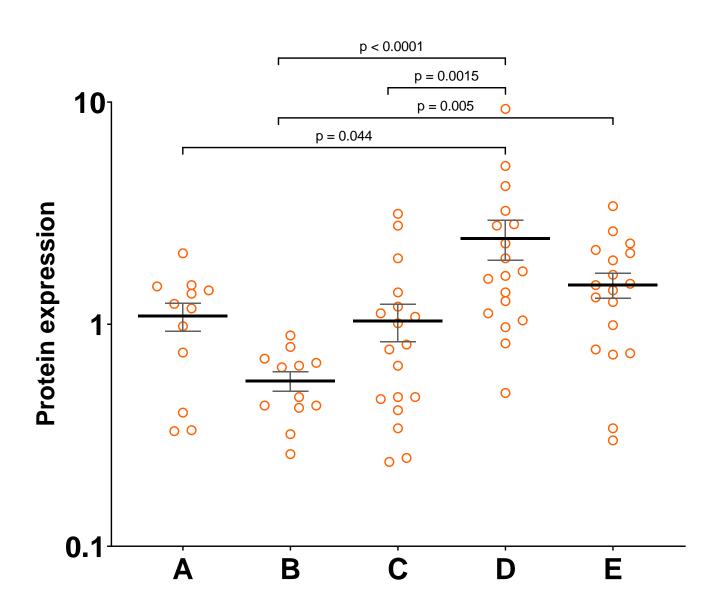
One-Way ANOVA in Prism 8



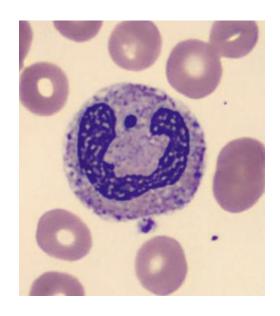
Have a go!



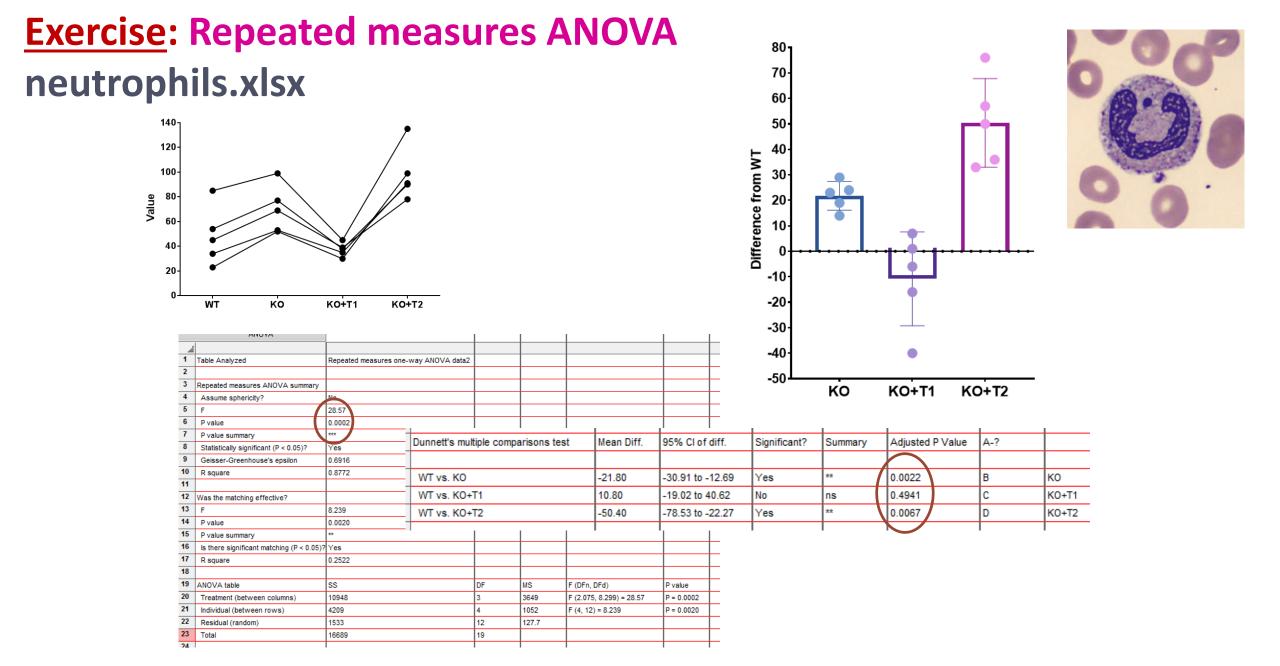
Analysis of variance: results



Exercise: Repeated measures ANOVA neutrophils.xlsx



- A researcher is looking at the difference between 4 cell groups. He has run the
 experiment 5 times. Within each experiment, he has neutrophils from a WT (control), a
 KO, a KO+Treatment 1 and a KO+Treatment2.
- Question: Is there a difference between KO with/without treatment and WT?



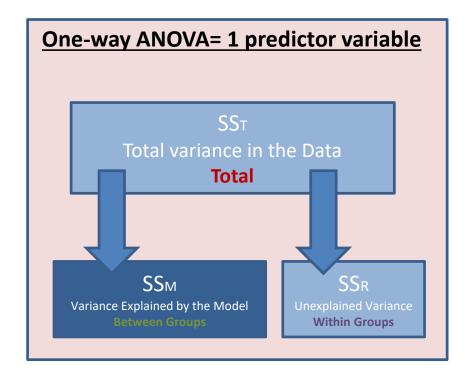
Answer: There is a significant difference from WT for the first and third groups.

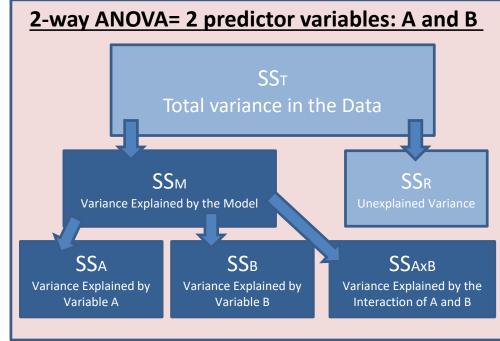
Comparison between more than 2 groups Two factors = Two predictors Two-Way ANOVA

Two-way Analysis of Variance (Factorial ANOVA)

Source of variation	Sum of	Df	Mean Square	F	p-value
	Squares				
Variable A (Between Groups)	2.665	4	0.6663	8.42	<0.0001
Within Groups (Residual)	5.775	73	0.0791		
Total	8.44	77			

Source of variation	Sum of Squares	Df	Mean Square	F	p-value
Variable A * Variable B	1978	2	989.1	F (2, 42) = 11.91	P < 0.0001
Variable B (Between groups)	3332	2	1666	F (2, 42) = 20.07	P < 0.0001
Variable A (Between groups)	168.8	1	168.8	F (1, 42) = 2.032	P = 0.1614
Residuals	3488	42	83.04		



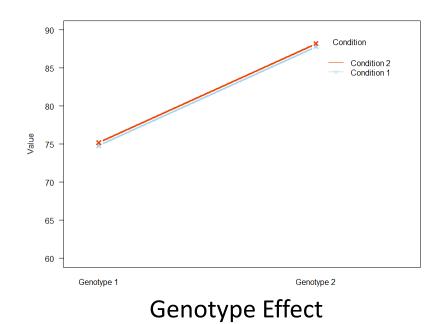


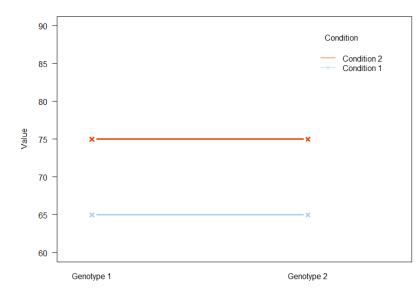
- Interaction plots: Examples
 - Fake dataset:
 - <u>2 factors</u>: **Genotype** (2 levels) and **Condition** (2 levels)

Genotype	Condition	Value
Genotype 1	Condition 1	74.8
Genotype 1	Condition 1	65
Genotype 1	Condition 1	74.8
Genotype 1	Condition 2	75.2
Genotype 1	Condition 2	75
Genotype 1	Condition 2	75.2
Genotype 2	Condition 1	87.8
Genotype 2	Condition 1	65
Genotype 2	Condition 1	74.8
Genotype 2	Condition 2	88.2
Genotype 2	Condition 2	75
Genotype 2	Condition 2	75.2

- Interaction plots: Examples
 - 2 factors: Genotype (2 levels) and Condition (2 levels)

Single Effect

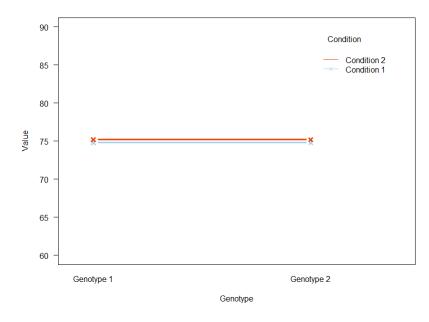


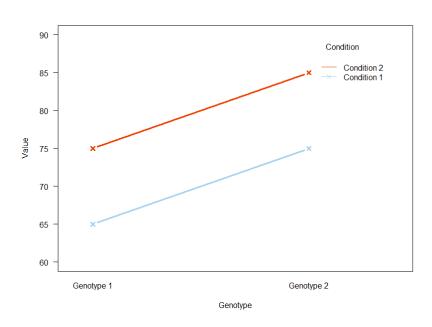


Condition Effect

- Interaction plots: Examples
 - <u>2 factors</u>: **Genotype** (2 levels) and **Condition** (2 levels)

Zero or Both Effect



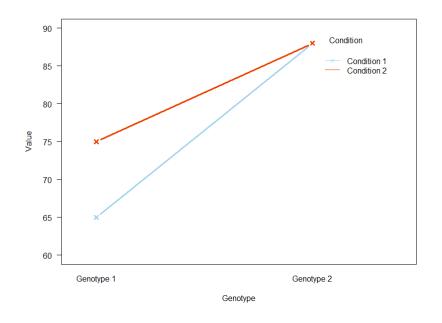


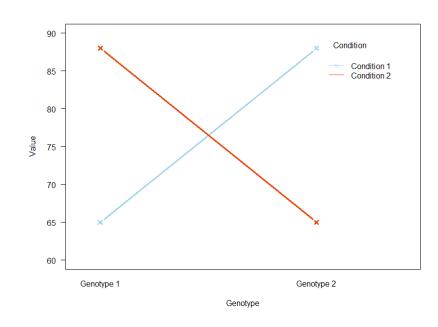
Zero Effect

Both Effect

- Interaction plots: Examples
 - 2 factors: Genotype (2 levels) and Condition (2 levels)

Interaction





Alcohol

Gender

Male

50

55

70

Female

70

55

2 Pints

Female

Male

55

50

4 Pints

30

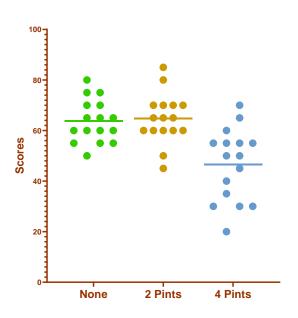
Female Male

60

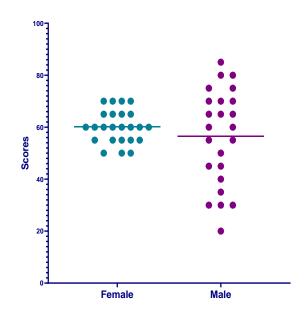
Example: goggles.xlsx

- The 'beer-goggle' effect
 - The term refers to finding people more attractive after you've had a few beers. Drinking beer provides a warm, friendly sensation, lowers your inhibitions, and helps you relax.
- Study: effects of alcohol on mate selection in night-clubs.
- Pool of independent judges scored the levels of attractiveness of the person that the participant was chatting up at the end of the evening.
- Question: is subjective perception of physical attractiveness affected by alcohol consumption?
 - Attractiveness on a scale from 0 to 100

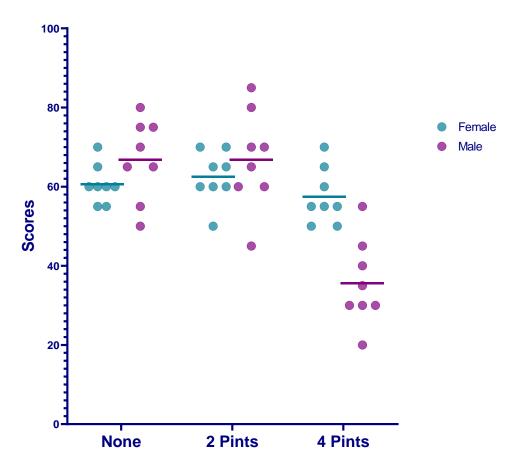
Main effect of Alcohol



Main effect of Gender



Interaction between Alcohol and Gender



With significant interaction (real data)

ANOVA table	SS	DF	MS	F (DFn, DFd)	P value
Interaction	1978	2	989.1	F (2, 42) = 11.91	< 0.0001
Alcohol Consumption	3332	2	1666	F (2, 42) = 20.07	< 0.0001
Gender	168.8	1	168.8	F (1, 42) = 2.032	0.1614
Residual	3488	42	83 04		

Without significant interaction (fake data)

ANOVA table	SS	DF MS	F (DFn, DFd)	P value
Interaction	7.292	2 3.646	F (2, 42) = 0.06872	0.9337
Alcohol Consumption	5026	2 2513	F (2, 42) = 47.37	< 0.0001
Gender	438.0	1 438.0	F (1, 42) = 8.257	0.0063
Residual	2228	42 53 05		

