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

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# Comparison between more than $\mathbf{2}$ groups <br> One factor = One predictor One-Way ANOVA 

## Signal-to-noise ratio



Signal = statistical significance Noise
$\underline{\text { Signal }}=$ no statistical significance
Noise

## Analysis of variance: how does it work?

$\underline{\text { Signal }}=\underline{\text { Difference between the means }}$
Noise $\quad$ Variability in the groups
$=F$ 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.


## Analysis of variance: how does it work?



| 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 |  |  |  |  |

## Analysis of variance: how does it work?



## Analysis of variance: how does it work?



5 differences: $\sum_{1}^{5}\left(\text { mean }_{n}-\text { grand mean }\right)^{2}$
$=$
Sum of squared errors
Between the groups

| Source of variation | Sum of Squares | df | Mean Square | F | p-value |
| :---: | :---: | :---: | :---: | :---: | :---: |
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| Within Groups |  |  |  |  |  |
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## Analysis of variance: how does it work?



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

$d f$ : degree of freedom with $d f=n-1$
$n=$ number of values, $k=n u m b e r$ of groups
Between groups: df = $4(k-1)$
Within groups: $d f=73\left(n-k=n_{1}-1+\ldots+n_{5}-1\right)$

## Analysis of variance: how does it work?



| Source of variation | Sum of Squares | df | Mean Squares | F ratio | p-value |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Between Groups | 18.1 | 4 | 4.5 |  |  |
|  | Within Groups | 51.8 | 73 | 0.71 |  |  |
|  | Total | 69.9 |  |  |  |  |

df: degree of freedom with $\mathrm{df}=\mathrm{n}-1$

$$
18.2 / 4=4.5 \quad 51.8 / 73=0.71
$$

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

## Analysis of variance: how does it work?



| Source of variation | Sum of Squares | df | Mean Squares | F ratio | p-value |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Between Groups | 18.1 | 4 | 4.5 | 6.34 | 0.0002 |
| Within Groups | 51.8 | 73 | 0.71 |  |  |
| Total | 69.9 |  |  |  |  |

Mean squares $=$ Sum of Squares $/ \mathbf{n - 1}=$ Variance
F ratio $=\frac{\text { Variance between the groups }}{\text { Variance within the groups (individual variability) }}=\frac{4.5}{0.71}=6.34$

## 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 not making the Type I Error 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 $\mathbf{1 4 . 3 \%}$
- The probability has increased from $5 \%$ to $14.3 \%$
- Comparisons between 5 groups instead of 3 , the familywise error rate is $\mathbf{4 0 \%}\left(=1-(0.95)^{\mathrm{n}}\right)$


## 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: $\alpha_{\text {adjust }}=0.05 / \mathrm{n}$ comparisons e.g. 3 comparisons: $0.05 / 3=0.016$
- Problem: very conservative leading to loss of power (lots of false negative)
- 10 comparisons: threshold for significance: 0.05/10: 0.005
- Pairwise comparisons across 20.000 genes $)$
- FDR: 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 at least 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 |
| 2 |  |  |  |  | - |  |
| 26 | Number of values | 12 | 12 | 18 | 18 | 18 |
| -- |  |  |  |  |  |  |





Transform of Protein expression


## Parametric tests assumptions

| 4 |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 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 |
| 25 |  |  |  |  |  |  |
| 26 | Number of values | 12 | 12 | 18 | 18 | 18 |



## One-Way ANOVA in Prism 8




## 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?


## Exercise: Repeated measures ANOVA

 neutrophils.xlsx



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

Comparison between more than $\mathbf{2}$ groups Two factors = Two predictors Two-Way ANOVA

## Two-way Analysis of Variance (Factorial ANOVA)

| Source of variation | Sum of <br> Squares | Df | Mean Square | F | p-value |
| :--- | :--- | :--- | :--- | :--- | :--- |
| 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 |  |  |  |

## One-way ANOVA=1 predictor variable



| 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 |  |  |



## Two-way Analysis of Variance

## - Interaction plots: Examples

- Fake dataset:
- 2 factors: 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 |

## Two-way Analysis of Variance

- Interaction plots: Examples
- $\underline{2 \text { factors: Genotype (2 levels) and Condition (2 levels) }}$


## Single Effect




## Two-way Analysis of Variance

- Interaction plots: Examples
- $\underline{2 \text { factors: Genotype (2 levels) and Condition (2 levels) }}$

Zero or Both Effect


Zero Effect


Both Effect

## Two-way Analysis of Variance

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

Interaction



## Two-way Analysis of Variance

## Example: goggles.xlsx

- The 'beer-goggle' effect

| Alcohol | None |  |  | 2 Pints |  | 4 Pints |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :---: |
| Gender | Female | Male | Female | Male | Female | Male |  |
| 65 | 50 | 70 | 55 | 45 | 30 |  |  |
| 70 | 55 | 65 | 65 | 60 | 30 |  |  |
| 60 | 80 | 60 | 70 | 85 | 30 |  |  |
| 60 | 65 | 70 | 55 | 65 | 55 |  |  |
| 60 | 70 | 65 | 55 | 70 | 35 |  |  |
|  | 75 | 75 | 60 | 60 | 70 | 20 |  |
|  | 60 | 75 | 60 | 50 | 80 | 45 |  |
|  | 65 | 65 | 50 | 50 | 60 | 40 |  |

- 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


## Two-way Analysis of Variance



## Two-way Analysis of Variance

## With significant interaction (real data)

| ANOVA table | SS DF |  | MS | $F(D F n, ~ D F d)$ | P value |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Interaction | 1978 | 2989.1 | $F(2,42)=11.91$ | $<0.0001$ |  |
| Alcohol Consumption | 3332 | 21666 | $F(2,42)=20.07$ | $<0.0001$ |  |
| Gender | 168.8 | 1168.8 | $F(1,42)=2.032$ | 0.1614 |  |
| Residual | 3488 | 4283.04 |  |  |  |



## Without significant interaction (fake data)

| ANOVA table | SS DF | MS | $F(D F n, D F d)$ | P value |
| :--- | ---: | ---: | ---: | ---: |
| Interaction | 7.292 | 23.646 | $F(2,42)=0.06872$ | 0.9337 |
| Alcohol Consumption | 5026 | 22513 | $F(2,42)=47.37$ | $<0.0001$ |
| Gender | 438.0 | 1438.0 | $F(1,42)=8.257$ | 0.0063 |
| Residual | 2228 | 4253.05 |  |  |

## Two-way Analysis of Variance

## Analyze Data

Built-in analysis
$\boxminus$ Transform, Normalize. Transform
Transform concentrations $(x)$
Normalize
Prune rows
Remove baseline and column math
Transpose $X$ and $Y$
Fraction of total

## © XY analyses

© Column analyses
$\boxminus$ Grouped analyses
TWO-Way ANOVA (or mixed model)
Three-way ANOVA (or mixed model)
Row means with SD or SEM
Multiple $t$ tests - one per row (1) Contingency table analyses $\boxplus$ Survival analyses
© Parts of whole analyses
© Multiple variable analyses
$\boxplus$ Nested analyses
© Generate curve
$\boxplus$ Simulate data

Parameters: Two-Way ANOVA (or Mixed Model) $\times$
RM Design RM Analysis Factor names Multiple Comparisons Options Residuals Multiple comparisons test
O Correct for multiple comparisons using statistical hypothesis testing. Recommended. Iest: Slidak (more power, recommended)
O Correct for multiple comparisons by controlling the Ealse Discovery Rate.
Test: Two-stage step-up method of Berjiamini, Krieger and Yelutieli (recommended)
O Don't correct for multiple comparisons. Each comparison stands alone.
Multiple comparisons options
$\square$ swnap direction of comparisons ( $^{(A-B)}$ vs. ( $(B-A)$.
$\square$ Report multiplicty adiusted P value for each comparison.
Each $P$ value is adjusted to account for multiple comparisons.
Family-wise significance and conffience level: 0.05 ( $95 \%$ confidence interval) $\vee$
Graphing options Graphing options
$\square$ Graph confidence intervals.
Additional results
$\square$ Narrative results.
$\square$ Narrative Eesults.
$\square$ Show cell/row/column/grand means.
Report goodness of fit.
Output
Show this many significant digits (for everything except P values): $\quad 4 \mid-$
P yalue style: GP: $0.1234(\mathrm{~ns}), 0.0332\left({ }^{( }\right), 0.0021(*), 0 . c \vee \mathbb{N}=6 \div$
$\square$ Make options on this tab be the defaul for future Two-Way ANOVAs.

Have a go!

## Two-way Analysis of Variance



