

# Analysis of variance (ANOVA)

(STATS201.stats 201 30: Statistical inference)

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# Introduction

- In lecture 110 we discussed inference on the mean of a population ( $\mu$ ).
- In the former part of lecture 120 we discussed inference on the difference of the mean of two populations ( $\mu_1 - \mu_2$ ).
- In the latter part of lecture 120 we discussed inference on paired measurements ( $\mu_D$ ).
- Now we will discuss a method that we can apply to compare the means of two or more populations. The method is called analysis of variance, or ANOVA.

## Analysis of variance

- Analysis of variance is one of the most commonly used statistical methods. There are several variants of it that can be used in a vast number of various different cases.
- We will only look at one variant of the method that is called *one-sided ANOVA*.
- It is applied to data that contain samples from two or more populations and it is common to speak of groups when discussing the samples.
- The method compares the variability of the measurements within the groups on one hand and between them on the other hand.
- ANOVA assumes that the samples are random samples, that they are sampled from populations with a normal distribution and that

# Conducting hypothesis tests

## Conducting hypothesis tests

- 1 Decide which hypothesis test is appropriate for our measurements.
- 2 Decide the  $\alpha$ -level.
- 3 Propose a null hypothesis and decide the direction of the test (one- or two-sided).
- 4 Calculate the test statistic for the hypothesis test.
- 5a See whether the test statistic falls within the rejection interval.
- 5b Look at the p-value of the test statistic.
- 6 Draw conclusions.

Average change group 1: 1. = 8.14  
Average change group 2: 2. = 6.28  
Average change group 3: 3. = 13.01

question is, do the drug decrease the blood pressure equally or not?

# The data

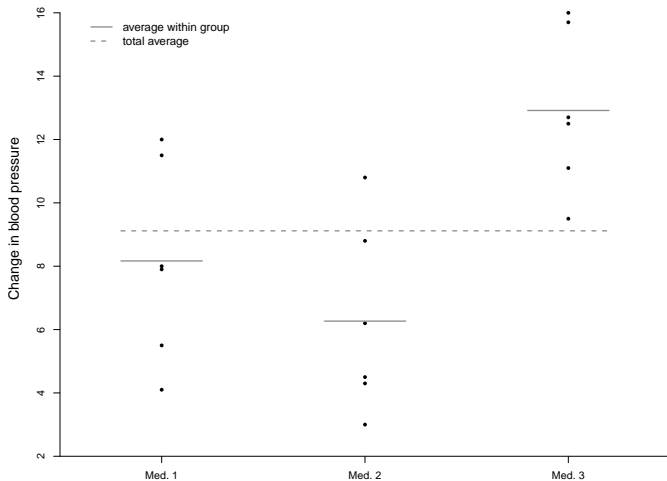


Figure: Data for ANOVA

# Syntax

## Syntax in ANOVA

The following syntax is common in textbooks and papers discussing ANOVA.

$y_{ij}$  :  $i$  denotes the number of the group and  $j$  denotes the number of a measurement within a group.  $y_{ij}$  is the  $j$ -th measurement in group  $i$ .

$a$  : We denote the number of groups with  $a$ .

$n_i$  : We denote the number of measurements in group  $i$  with  $n_i$ .

$N$  : The total number of measurements is denoted with  $N$ .

$$N = n_1 + n_2 + \dots + n_a.$$

$\bar{y}_i$  :  $\bar{y}_i$  denotes the mean of group  $i$

$$\bar{y}_i = \frac{\sum_{j=1}^{n_i} y_{ij}}{n_i}.$$

$\bar{y}_{..}$  :  $\bar{y}_{..}$  denotes the overall mean of all measurements (in all groups).

$$\bar{y}_{..} = \frac{\sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}}{N}.$$

## Sums of squares

- We need to calculate three sums of squares, and are they denoted with  $SS_T$ ,  $SS_{Tr}$  and  $SS_E$ .
- $SS_T$  is the total sums of squares and is a measure of the total variation of the measurements.
- $SS_{Tr}$  is a measure of the variation between groups (or treatments), that is, how much to the means of the groups vary.
- $SS_E$  is a measure of the variability within groups (or treatments) and is therefore a measure of the error. It shows how much the measurements deviate from the mean of the group.



# Sums of squares

## Sums of squares in one sided ANOVA

The Sums of squares are calculated with

$$SS_T = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2$$

$$SS_{Tr} = \sum_{i=1}^a n_i (\bar{y}_{i.} - \bar{y}_{..})^2$$

$$SS_E = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2$$

The total variation can be divided into the variation between groups on one hand and the variation within groups on the other hand or

$$SS_T = SS_{Tr} + SS_E.$$

# Sums of squares - graphically

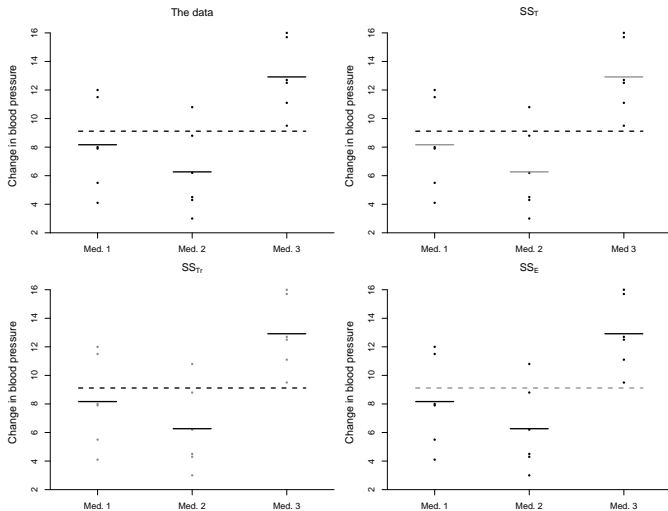


Figure: Sums of squares

## ANOVA table

- It is common to visualize the sums of squares in a so-called *ANOVA table*.
- The table consist of three columns and three lines.
- The first column contains the sums of squares, the next one contains the number of *degrees of freedom*. The first column contains the sums of squares, the second one contains the number of *degrees of freedom* for each sum of squares and the third column contains so-called mean sum of squares.
- Mean sum of squares is calculated by dividing the corresponding sum of squares with the number of corresponding degrees of freedom (in the same line).

## ANOVA table

Sums of squares	Degrees of freedom	Mean sum of squares
$SS_{Tr}$	$a - 1$	$MS_{Tr} = \frac{SS_{Tr}}{a-1}$
$SS_E$	$N - a$	$MS_E = \frac{SS_E}{N-a}$
$SS_T$	$N - 1$	

# Hypothesis testing with ANOVA

## Hypothesis testing with ANOVA

The hypothesis we want to test is generally

$$H_0 : \mu_1 = \mu_2 = \dots = \mu_a$$

against the alternative hypothesis

$H_1$  : At least one of the means differs from the other means.

The test statistic is

$$F = \frac{SS_{Tr}/(a-1)}{SS_E/(N-a)} = \frac{MS_{Tr}}{MS_E}.$$

If the null hypothesis is true, the test statistic follows the F-distribution with  $a-1$  and  $N-a$  degrees of freedom, or  $F \sim F_{(a-1, N-a)}$ , where  $a$  is the number of groups and  $N$  is the total number of measurements.

$H_0$  is rejected if  $F > F_{1-\alpha, (a-1, N-a)}$

# Hypothesis testing with one-sided ANOVA

- The alternative hypothesis is that at least one of the means differs from the others, it is therefore then only information we receive if the null hypothesis is rejected.
- We do not know which of the means differs from the others or if they are potentially all different.
- Further analysis needs to be done in order to find that out. A common test is Tukey's test, but they will not be covered in this lecture.