# **Statistics Kingdom**

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## **ANOVA Calculator**

One-Way ANOVA Calculator and Tukey HSD

Significance level (a):

Outliers:

Included

Fffect:

Effect type:

Medium

Medium

Ff

Rounding:

0.25

Tutorial ANOVA

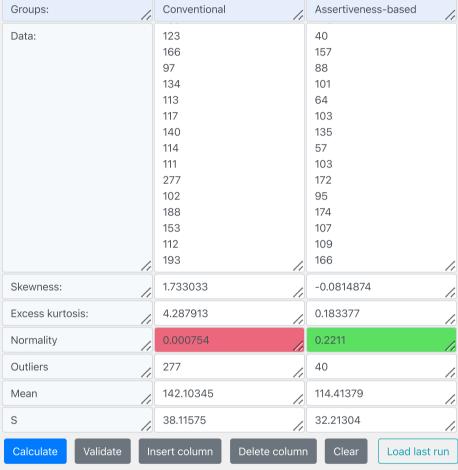
Calculators

Kruskal-Wallis test
Two way ANOVA
Levene's test

Enter raw data directly

O Enter raw data from excel

# Enter sample data directly



**Header**: You may change groups' names to the real names.

 $\textbf{Data} \hbox{: When entering data, press } \boxed{ \texttt{Enter} } \hbox{ or } \boxed{ , } \hbox{ (comma) after each value}.$ 

You may paste full column from excel.

The calculator ignores empty cells or non-numeric cells.

Group1 contains 29 values Group2 contains 29 values

validation:success

Hover over the cells for more information.

Source	DF	Sum of Square	Mean Square	F Statistic	P-value
Groups (between groups)	1	11117.3964	11117.3964	8.9279	0.004163
Error (within groups)	56	69733.7239	1245.2451		
Total	57	80851.1203	1418.4407		

#### R code.







#### One Way ANOVA test, using F distribution df(1,56) (right tailed)

### <u>1. H<sub>0</sub> hypothesis</u>

Since p-value< $\alpha$ ,  $H_0$  is rejected.

Some of the groups' averages consider to be not equal.

In other words, the difference between the averages of some groups is big enough to be statistically significant.

#### 2. P-value

p-value equals 0.00416348, [p( x  $\leq$  F ) = 0.995837 ]. It means that the chance of type1 error (rejecting a correct H<sub>0</sub>) is small: 0.004163 (0.42%) The smaller the p-value the stronger it support H<sub>1</sub>

#### 3. The statistics

The test statistic F equals **8.927878**, which is not in the 95% region of acceptance:  $[-\infty : 4.013]$ 

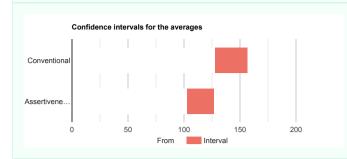
#### 4. Effect size

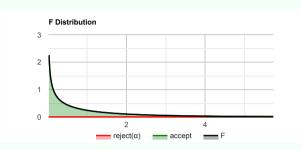
The observed effect size f is large (0.4). That indicates that the magnitude of the difference between the averages is large.

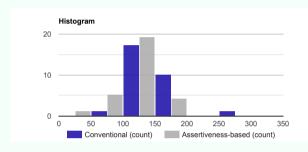
The  $\eta^2$  equals 0.14. It means that the **group** explains 13.8% of the variance from the average (similar to  $R^2$  in the linear regression)

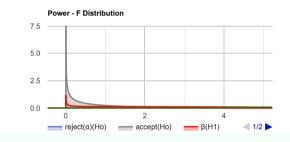
#### 5. Tukey HSD / Tukey Kramer

The means of the following pair are significantly different: x1-x2.









# <u>Validation</u>

#### Test power

Although the priori power is low (0.4647), the  $\mbox{\rm H}_{0}$  is rejected.

#### Equality of variances

The tool used the Levene's test to assess the equality of variances.

The population's variances consider to be **equal**. (p-value = 0.497).

Levene's test power consider to be weak (0.46).

The groups' size consider similar. (The ratio between the bigger group and the smaller group is: 1)

The ANOVA test consider to be robust to the homogeneity of variances assumtion when the groups' sizes are similar.

It is suggested to consider the  ${\bf Kruskal\text{-}Wallis}$   ${\bf ANOVA.}$  non-parametric test.

#### Normality assumption

The assumption was checked based on the <u>Shapiro-Wilk Test</u>. ( $\alpha$ =0.05)

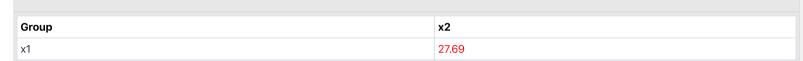
The ANOVA test is considered robust for moderate violation of the normality assumption.

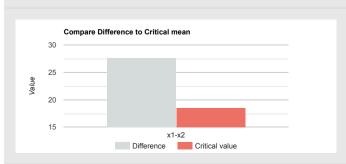
From the groups with small sample size, less than 30, **1** group doesn't distribute normally (p-value is 0.000754).

You should consider transform the data or use the Kruskal–Wallis non-parametric test.

# **Tukey HSD / Tukey Kramer**

Pair	Difference	SE	Q	Lower CI	Upper CI	Critical Mean	p-value
x1-x2	27.6897	6.5528	4.2256	9.1254	46.2539	18.5642	0.004163





### **ANOVA**

#### What is the ANOVA?

The ANOVA test checks if the difference between the averages of two or more groups is significant, using sample data.

ANOVA is usually used when there are at least three groups since for two groups, the *two-tailed* pooled variance t-test and the *right-tailed* ANOVA test have the same result.

The basic ANOVA test contains only one categorical value, one-way ANOVA. For example, if you compare the performence of three schools, the categorical variable is school, and the possible values of the categorical variable are School-A, School-B, School-C. There are more complex ANOVA tests that contain two categorical variables (<a href="Two-way ANOVA calculator">Two-way ANOVA calculator</a>), or more. When performing a one-way ANOVA test, we try to determine if the difference between the averages reflects a real difference between the groups, or is due to the random noise inside each group.

The F statistic represents the ratio of the variance between the groups and the variance inside the groups. Unlike many other statistic tests, the smaller the F statistic the more likely the averages are equal.

**Example**: Compare four fertilizers used in four fields

 $H_0$ : The average weight of crops per square meter is equal in all fields.

H<sub>1</sub>: At least one field yields a different average per square meter.

Right-tailed, for ANOVA test you can use only the right tail. Why?



 $H_1$ : not( $\mu_1 = ... = \mu_k$ )

# ANOVA formula

$$F = \frac{MSO}{MSF}$$



# **Assumptions**

- Independent samples
- Normal distribution of the analyzed population
- Equal standard deviation,  $\sigma_1 = \sigma_2 = ... = \sigma_k$

The assumption is more important when the groups' sizes not similar

# **Required Sample Data**

Sample data from all compared groups

#### **Parameters**

- k Number of groups.
- n<sub>i</sub> Sample side of group i
- ${\bf n}$  Overall sample side, includes all the groups ( $\Sigma n_i$ , i=1 to k)
- $\mathbf{x}_{i}$  Average of group i.
- $\vec{x}$  Overall average ( $\sum x_{i,j} / n$ , i=1 to k, j=1 to  $n_i$ )
- Si Standard deviation of group i

## Results calculations

Source	Degrees of Freedom	Sum of Squares	Mean Square	F statistic	p-value
Groups (between groups)	k - 1	$SSG = \sum_{i=1}^{k} n_i (\bar{x}_i - \bar{x})^2$	MSG = SSG / (k - 1)	F = MSG / MSE	P(x > F)
Error (within groups)	n - k	$SSE=\sum_{i=1}^{k}(n_i-1)S_i^2$	MSE = SSE / (n - k)		
Total	n - 1	SS(total) = SSG + SSE	Sample Variance = SS(total) / (n - 1)		

### R Code

The following R code should produce the same results

#### value<

my.dataframe<-data.frame(value, group1)

res.aov <- aov(value ~ group1, data = my.dataframe)

summary(res.aov)

TukeyHSD(res.aov)