

# Statistics Kingdom

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## ANOVA Calculator

One-Way ANOVA Calculator and Tukey HSD

[Tutorial](#)  
[ANOVA](#)  
**Calculators**  
[Kruskal-Wallis test](#)  
[Two way ANOVA](#)  
[Levene's test](#)

Significance level ( $\alpha$ ):

0,05

Outliers:

Included

Effect:

Medium

Effect type:

f

Effect Size:

0.25

Rounding:

4

- ☒ Enter raw data directly
- ☐ Enter raw data from excel

## Enter sample data directly

Groups:	Conventional	Assertiveness-based
Data:	5 5 4 4 4 4 4 5 4 5 5 5 5 5 5	5 5 5 4 4 4 4 5 4 5 5 4 5 5 5
Skewness:	-1.100391	-0.684484
Excess kurtosis:	0.193622	-1.686667
Normality	0.00001476	0.000001046
Outliers		
Mean	4.52174	4.65217
S	0.66535	0.48699

Calculate

Validate

Insert column

Delete column

Clear

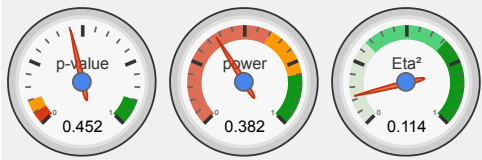
Load last run

**Header:** You may change groups' names to the real names.  
**Data:** When entering data, press `Enter` or `,` (comma) after each value.  
You may paste full column from excel.  
*The calculator ignores empty cells or non-numeric cells.*  
  
Group1 contains 23 values  
Group2 contains 23 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	0.1957	0.1957	0.5756	0.4521
Error (within groups)	44	14.9565	0.3399		
Total	45	15.1522	0.3367		

[R code.](#)



**One Way ANOVA test, using F distribution df(1,44) (right tailed)**

**1. H<sub>0</sub> hypothesis**

Since p-value > α, H<sub>0</sub> is accepted.  
The averages of all groups assumed to be equal.  
In other words, the difference between the averages of all groups is not big enough to be statistically significant.

**2. P-value**

p-value equals **0.452094**, [p( x ≤ F ) = 0.547906 ]. It means that if we would reject H<sub>0</sub>, the chance of type1 error (rejecting a correct H<sub>0</sub>) would be too high: 0.4521 (45.21%)  
The bigger the p-value the stronger it supports H<sub>0</sub>

**3. The statistics**

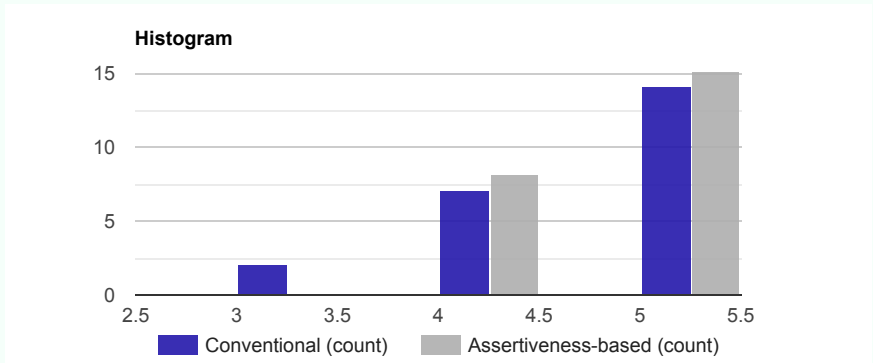
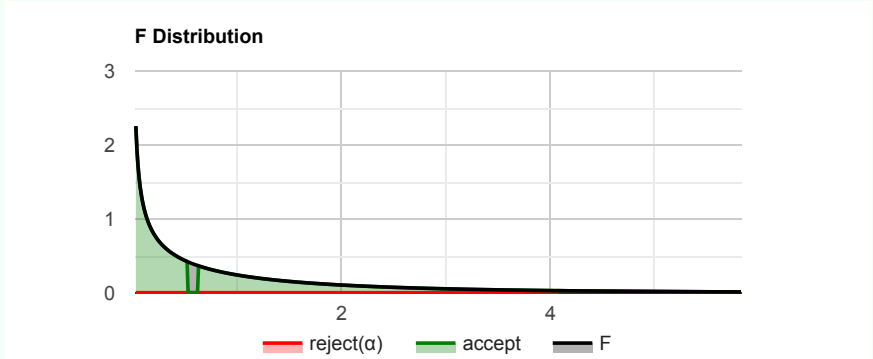
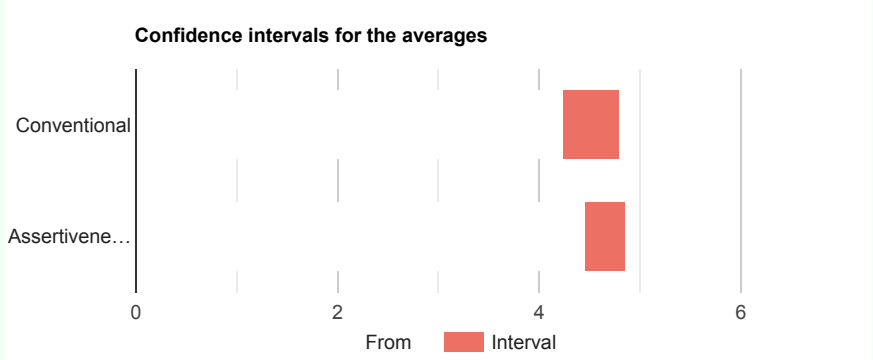
The test statistic F equals **0.575583**, which is in the 95% region of acceptance: [-∞ : 4.0617]

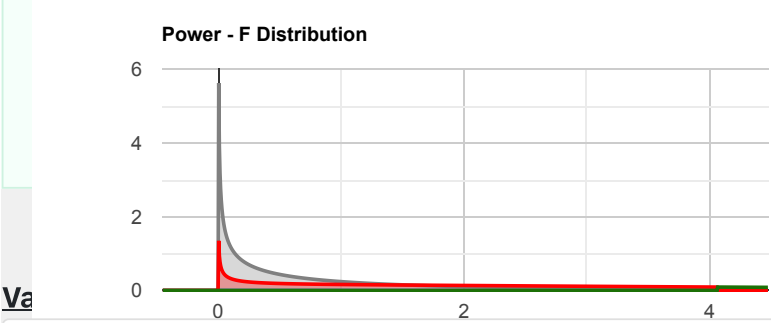
**4. Effect size**

The observed effect size f is **small** (0.11). That indicates that the magnitude of the difference between the averages is small.  
The η² equals 0.013. It means that the **group** explains 1.3% of the variance from the average (similar to R² in the linear regression)

**5. Tukey HSD / Tukey Kramer**

There is no significant difference between the means of any pair.





Value

● Test power

The priori power is low (0.3817), hence the test may not reject an incorrect  $h_0$

It is suggested to improve the test power by:

- **sample size:** use a larger sample.
- **$\sigma$ :** check if the standard deviation can be reduced by eliminating noises that are not relevant to the tested measurement.
- **effect size\*:** when planning the research it was possible to increase the effect size, at the price of the ability to identify smaller effect sizes.
- **$\alpha$ \*:** when planning the research it was possible to increase the significance level ( $\alpha$ ), at the price of increasing the probability of a type I error.

\*Note: determining the test power, sample size, effect size and the significant level ( $\alpha$ ) should be done **before** collecting the data.

● 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 = 0.452).  
Levene's test power consider to be weak (0.38).  
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 assumption when the groups' sizes are similar.

It is suggested to consider the **Kruskal-Wallis ANOVA**. non-parametric test.

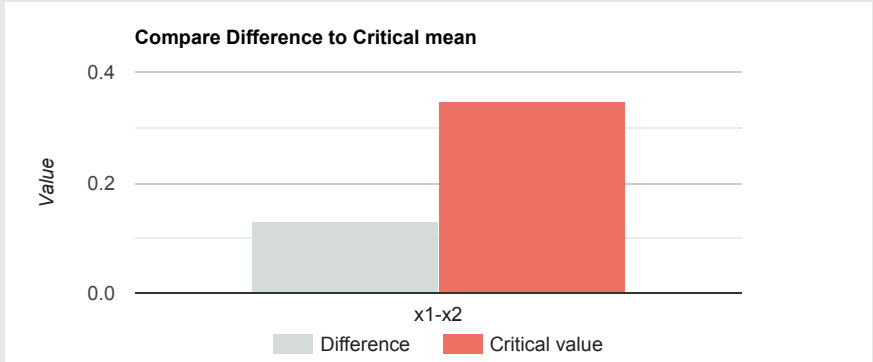
● Normality assumption

The assumption was checked based on the [Shapiro-Wilk Test](#). ( $\alpha=0.05$ )  
The ANOVA test is considered robust for moderate violation of the normality assumption.  
From the groups with small sample size, less than 30, **2** groups doesn't distribute normally (the smaller p-value is 0.00000105).  
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	0.1304	0.1216	1.0729	-0.2161	0.4769	0.3465	0.4521

Group	x2
x1	0.13



## 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 performance 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 ( [Two-way ANOVA calculator](#) ), 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<sub>0</sub>: 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?](#)

### Hypotheses

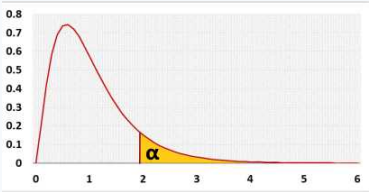
H<sub>0</sub>: μ<sub>1</sub> = ... = μ<sub>k</sub>

H<sub>1</sub>: not(μ<sub>1</sub> = ... = μ<sub>k</sub>)

### ANOVA formula

$$F = \frac{MSG}{MSE}$$

### F distribution



## Assumptions

- Independent samples
  - Normal distribution of the analyzed population
  - Equal standard deviation, σ<sub>1</sub>=σ<sub>2</sub>=...=σ<sub>k</sub>
- 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
- n** - Overall sample side, includes all the groups (Σn<sub>i</sub>, i=1 to k)
- $\bar{x}_i$**  - Average of group i.
- $\bar{x}$**  - Overall average (Σx<sub>i,j</sub> / n, i=1 to k, j=1 to n<sub>i</sub>)
- S<sub>i</sub>** - Standard deviation of group i

## Results calculations

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

## R Code

The following R code should produce the same results

```
value<-c(3,3,4,4,4,4,4,4,4,5,5,5,5,5,5,5,5,5,5,4,4,4,4,4,4,4,5,5,5,5,5,5,5,5,5,5,5,5)
group1 <- c(rep("Conventional", 23), rep("Assertiveness-based", 23))
my.dataframe<-data.frame(value, group1)
res.aov <- aov(value ~ group1, data = my.dataframe)
```

