

# Package ‘visStatistics’

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**Type** Package

**Title** Automated Selection and Visualisation of Statistical Hypothesis Tests

**Version** 0.1.7

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**Description** Automatically selects and visualises statistical hypothesis tests between two vectors, based on their class, distribution, sample size, and a user-defined confidence level (`conf.level`). Visual outputs - including box plots, bar charts, regression lines with confidence bands, mosaic plots, residual plots, and Q-Q plots - are annotated with relevant test statistics, assumption checks, and post-hoc analyses where applicable. The algorithmic workflow helps the user focus on the interpretation of test results rather than test selection. It is particularly suited for quick data analysis, e.g., in statistical consulting projects or educational settings. The test selection algorithm proceeds as follows:  
Input vectors of class `numeric` or `integer` are considered numerical; those of class `factor` are considered categorical. Assumptions of residual normality and homogeneity of variances are considered met if the corresponding test yields a p-value greater than the significance level  $\alpha = 1 - \text{conf.level}$ .  
(1) When the response vector is numerical and the predictor vector is categorical, a test of central tendencies is selected. If the categorical predictor has exactly two levels, `t.test()` is applied when group sizes exceed 30 (Lumley et al. (2002) <[doi:10.1146/annurev.publhealth.23.100901.140546](https://doi.org/10.1146/annurev.publhealth.23.100901.140546)>). For smaller samples, normality of residuals is tested using `shapiro.test()`; if met, `t.test()` is used; otherwise, `wilcox.test()`.  
If the predictor is categorical with more than two levels, an `aov()` is initially fitted. Residual normality is evaluated using both `shapiro.test()` and `ad.test()`; residuals are considered approximately normal if at least one test yields a p-value above  $\alpha$ . If this assumption is met, `bartlett.test()` assesses variance homogeneity. If variances are homogeneous, `aov()` is used; otherwise `oneway.test()`. Both tests are followed by `TukeyHSD()`.

If residual normality cannot be assumed, `kruskal.test()` is followed by `pairwise.wilcox.test()`.

(2) When both the response and predictor vectors are numerical, a simple linear regression model is fitted using `lm()`.

(3) When both vectors are categorical, Cochran's rule (Cochran (1954) <[doi:10.2307/3001666](https://doi.org/10.2307/3001666)>)

is applied to test independence either by `chisq.test()` or `fisher.test()`.

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**URL** <https://github.com/shhschilling/visStatistics>,  
<https://shhschilling.github.io/visStatistics/>

**BugReports** <https://github.com/shhschilling/visStatistics/issues>

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counts_to_cases	<i>Convert data frame of counts to data frame of cases. data frame must contain a column with frequencies (counts) as generated by as.data.frame from a contingency table</i>
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**Description**

Convert data frame of counts to data frame of cases. data frame must contain a column with frequencies (counts) as generated by as.data.frame from a contingency table

**Usage**

```
counts_to_cases(x, countcol = "Freq")
```

**Arguments**

x	a data.frame of counts generated from a contingency table.
countcol	character string, name of the column of x containing the counts. Default name of the column is 'Freq'.

**Value**

data frame of cases of dimension (total number of counts as sum of 'Freq' in x) times 2.

**Examples**

```
counts_to_cases(as.data.frame(HairEyeColor[, , 1]), countcol = "Freq")
```

---

openGraphCairo	<i>Cairo wrapper function</i>
----------------	-------------------------------

---

**Description**

Cairo wrapper function returning NULL if not type is specified

**Usage**

```
openGraphCairo(
  width = 640,
  height = 480,
  fileName = NULL,
  type = NULL,
  fileDirectory = getwd(),
  pointsize = 12,
  bg = "transparent",
```

```

    canvas = "white",
    units = "px",
    dpi = 150
  )

```

### Arguments

width	see Cairo()
height	see Cairo()
fileName	name of file to be created. Does not include both file extension <code>'type'</code> and file directory. Default file name <code>'visstat_plot'</code> .
type	Supported output types are <code>'png'</code> , <code>'jpeg'</code> , <code>'pdf'</code> , <code>'svg'</code> , <code>'ps'</code> and <code>'tiff'</code> . See Cairo()
fileDirectory	path of directory, where plot is stored. Default current working directory.
pointsize	see Cairo()
bg	see Cairo()
canvas	see Cairo()
units	see Cairo()
dpi	DPI used for the conversion of units to pixels. Default value 150.

### Details

openGraphCairo() Cairo() wrapper function. Differences to Cairo: a) prematurely ends the function call to Cairo() returning NULL, if no output type of types `'png'`, `'jpeg'`, `'pdf'`, `'svg'`, `'ps'` or `'tiff'` is provided. b) The file argument of the underlying Cairo function is generated by `file.path(fileDirectory, paste(fileName, '.', type, sep = ''))`.

### Value

NULL, if no type is specified. Otherwise see Cairo()

### Examples

```

## adapted from example in \code{Cairo()}
openGraphCairo(fileName = "normal_dist", type = "pdf", fileDirectory = tempdir())
plot(rnorm(4000), rnorm(4000), col = "#ff000018", pch = 19, cex = 2)
dev.off() # creates a file 'normal_dist.pdf' in the directory specified in fileDirectory
# ## remove the plot from fileDirectory
file.remove(file.path(tempdir(), "normal_dist.pdf"))

```

---

`plot.visstat`*Report only saved plot files for visstat objects*

---

### Description

This method reports the file paths of plots that were saved to disk during the statistical analysis, if the user requested file output (e.g., PNG or PDF) via the `graphicsoutput` argument.

### Usage

```
## S3 method for class 'visstat'  
plot(x, ...)
```

### Arguments

<code>x</code>	An object of class "visstat", returned by <code>visstat()</code> .
<code>...</code>	Currently unused. Included for S3 method compatibility.

### Details

Note that plots are shown during execution of `visstat()`, even if no files are saved. This method does not re-render or replay those plots.

All file-based plots in `visStatistics` are generated using the `Cairo()` device and their file paths are collected in the object's "plot\_paths" attribute.

In interactive sessions, this method prints the list of saved files. In non-interactive contexts, it suppresses output.

The `visstat()` function may produce several plots per test type (e.g., main effect, post hoc comparisons, assumption checks). These are saved to disk, and their file paths are collected in the object's "plot\_paths" attribute.

In an interactive session, `plot()` will print these paths to the console. In non-interactive sessions, it quietly lists them. No plots are shown or rendered within R.

### Value

Invisibly returns `x`. Used for its side effect of reporting file paths.

### See Also

[visstat](#), [Cairo](#), [print.visstat](#), [summary.visstat](#)

print.visstat            *Print method for visstat objects*

---

### Description

Displays a brief summary of the statistical test results and, if available, assumption tests and post hoc comparisons.

### Usage

```
## S3 method for class 'visstat'  
print(x, ...)
```

### Arguments

x                    An object of class "visstat", returned by visstat().  
...                  Currently unused. Included for S3 method compatibility.

### Details

This function is automatically called when a visstat object is printed to the console. It provides a quick overview of the statistical analysis results.

### Value

The object x, invisibly.

### See Also

[summary.visstat](#), [plot.visstat](#), [visstat](#)

---

saveGraphVisstat        *Saves Graphical Output*

---

### Description

Closes all graphical devices with dev.off() and saves the output only if both fileName and type are provided.

### Usage

```
saveGraphVisstat(  
  fileName = NULL,  
  type = NULL,  
  fileDirectory = getwd(),  
  oldfile = NULL  
)
```

**Arguments**

fileName	name of file to be created in directory fileDirectory without file extension '.type'.
type	see Cairo().
fileDirectory	path of directory, where graphic is stored. Default setting current working directory.
oldfile	old file of same name to be overwritten

**Value**

NULL, if no type or fileName is provided, TRUE if graph is created

**Examples**

```
# very simple KDE (adapted from example in Cairo())
openGraphCairo(type = "png", fileDirectory = tempdir())
plot(rnorm(4000), rnorm(4000), col = "#ff000018", pch = 19, cex = 2)
# save file 'norm.png' in directory specified in fileDirectory
saveGraphVisstat("norm", type = "png", fileDirectory = tempdir())
file.remove(file.path(tempdir(), "norm.png")) # remove file 'norm.png'
```

---

summary.visstat

*Summary method for visstat objects*


---

**Description**

Displays the full statistical test results and, if available, assumption tests and post hoc comparisons.

**Usage**

```
## S3 method for class 'visstat'
summary(object, ...)
```

**Arguments**

object	An object of class "visstat", returned by visstat().
...	Currently unused. Included for S3 method compatibility.

**Details**

This method provides a full textual report of the statistical test results returned by visstat(), and prints the contents of posthoc\_summary if present.

**Value**

The object object, invisibly.

**See Also**

[print.visstat](#), [visstat](#)

---

visstat

*Wrapper for visstat\_core allowing two different input styles*

---

**Description**

A wrapper around the core function [visstat\\_core](#) defining the decision logic for statistical hypothesis testing and visualisation between two variables of class "numeric", "integer", or "factor".

**Usage**

```
visstat(
  x,
  y,
  ...,
  conf.level = 0.95,
  numbers = TRUE,
  minpercent = 0.05,
  graphicsoutput = NULL,
  plotName = NULL,
  plotDirectory = getwd()
)
```

**Arguments**

x	A vector of class "numeric", "integer", or "factor" (standardised usage), or a data.frame containing the relevant columns (backward-compatible usage).
y	A second vector (standardised usage), or a character string specifying the name of a column in x (backward-compatible usage).
...	If x is a data frame and y is a character string, an additional character string must follow, naming the second column.
conf.level	Confidence level for statistical inference; default is 0.95.
numbers	Logical. Whether to annotate plots with numeric values.
minpercent	Minimum proportion (between 0 and 1) required to display a category in plots.
graphicsoutput	Optional. Output format for plots (e.g., "pdf", "png").
plotName	Optional. File name prefix for saving plot output.
plotDirectory	Directory in which to save plots; defaults to the current working directory.

## Details

This wrapper supports two input formats:

- **Standardised form:** `visstat(x, y)`, where both `x` and `y` are vectors of class `"numeric"`, `"integer"`, or `"factor"`.
- **Backward-compatible form:** `visstat(dataframe, "name_of_y", "name_of_x")`, where both character strings refer to column names in `dataframe`. This is equivalent to: `visstat(dataframe[["name_of_x"]], dataframe[["name_of_y"]])`.

The interpretation of `x` and `y` depends on the variable classes: In the following, data of class `numeric` or `integer` are both referred to by their common mode `numeric`

- If one variable is `numeric` and the other a `factor`, the `numeric` vector must be passed as `y` and the `factor` as `x`. This supports tests of central tendencies (e.g., `t-test`, `ANOVA`, `Wilcoxon`).
- If both variables are `numeric`, a linear model is fitted with `y` as the response and `x` as the predictor.
- If both variables are `factors`, an association test (`Chi-squared` or `Fisher's exact`) is used. The test result is invariant to variable order, but visualisations (e.g., axis layout, bar orientation) depend on the roles of `x` and `y`.

This wrapper standardises the input and calls `visstat_core`, which selects and executes the appropriate test with visual output and assumption diagnostics.

## Value

A list as returned by `visstat_core`, containing statistical results and graphical outputs.

## See Also

the core function `visstat_core`, the package's vignette `vignette("visStatistics")` for the overview, and the accompanying webpage <https://shhschilling.github.io/visStatistics/>.

## Examples

```
## Standardised usage (preferred):
visstat(mtcars$am, mtcars$mpg)

## Backward-compatible usage (same result):
visstat(mtcars, "mpg", "am")

## Wilcoxon rank sum test
grades_gender <- data.frame(
  Sex = as.factor(c(rep("Girl", 20), rep("Boy", 20))),
  Grade = c(
    19.3, 18.1, 15.2, 18.3, 7.9, 6.2, 19.4, 20.3, 9.3, 11.3,
    18.2, 17.5, 10.2, 20.1, 13.3, 17.2, 15.1, 16.2, 17.3, 16.5,
    5.1, 15.3, 17.1, 14.8, 15.4, 14.4, 7.5, 15.5, 6.0, 17.4,
    7.3, 14.3, 13.5, 8.0, 19.5, 13.4, 17.9, 17.7, 16.4, 15.6
  )
)
```

```

visstat(grades_gender$Sex, grades_gender$Grade)

## Welch's one-way ANOVA
visstat(npk$block, npk$yield)

## Kruskal-Wallis
visstat(iris$Species, iris$Petal.Width)

## Simple linear regression
visstat(trees$Height, trees$Girth, conf.level = 0.99)

## Chi-squared
HairEyeColorDataFrame <- counts_to_cases(as.data.frame(HairEyeColor))
visstat(HairEyeColorDataFrame$Eye, HairEyeColorDataFrame$Hair)

## Fisher's test
HairEyeColorMaleFisher <- HairEyeColor[, , 1]
blackBrownHazelGreen <- HairEyeColorMaleFisher[1:2, 3:4]
blackBrownHazelGreen <- counts_to_cases(as.data.frame(blackBrownHazelGreen))
visstat(blackBrownHazelGreen$Eye, blackBrownHazelGreen$Hair)

## Save PNG
visstat(blackBrownHazelGreen$Hair, blackBrownHazelGreen$Eye,
        graphicsoutput = "png", plotDirectory = tempdir())

## Save PDF
visstat(iris$Species, iris$Petal.Width, graphicsoutput = "pdf",
        plotDirectory = tempdir())

## Custom plot name
visstat(iris$Species, iris$Petal.Width,
        graphicsoutput = "pdf", plotName = "kruskal_iris", plotDirectory = tempdir())

```

---

visstat\_core

*Automated Visualization of Statistical Hypothesis Testing*


---

## Description

`visstat_core()` provides automated selection and visualization of a statistical hypothesis test between a two vectors in a given `data.frame` named `dataframe` based on the data's type, distribution, sample size, and the specified `conf.level`. `varsample` and `varfactor` are character strings corresponding to the column names of the chosen vectors in `dataframe`. These vectors must be of type integer, numeric or factor. The automatically generated output figures illustrate the selected statistical hypothesis test, display the main test statistics, and include assumption checks and post hoc comparisons when applicable. The primary test results are returned as a list object.

## Usage

```
visstat_core(
```

```

dataframe,
varsample,
varfactor,
conf.level = 0.95,
numbers = TRUE,
minpercent = 0.05,
graphicsoutput = NULL,
plotName = NULL,
plotDirectory = getwd()
)

```

### Arguments

dataframe	data.frame with at least two columns.
varsample	character string matching a column name in dataframe. Interpreted as the response if the referenced column is of class numeric or integer and the column named by varfactor is of class factor.
varfactor	character string matching a column name in dataframe. Interpreted as the grouping variable if the referenced column is of class factor and the column named by varsample is of class numeric or integer.
conf.level	Confidence level
numbers	a logical indicating whether to show numbers in mosaic count plots.
minpercent	number between 0 and 1 indicating minimal fraction of total count data of a category to be displayed in mosaic count plots.
graphicsoutput	saves plot(s) of type "png", "jpg", "tiff" or "bmp" in directory specified in plotDirectory. If graphicsoutput=NULL, no plots are saved.
plotName	graphical output is stored following the naming convention "plotName.graphicsoutput" in plotDirectory. Without specifying this parameter, plotName is automatically generated following the convention "statisticalTestName_varsample_varfactor".
plotDirectory	specifies directory, where generated plots are stored. Default is current working directory.

### Details

The decision logic for selecting a statistical test is described below. For more details, please refer to the package's vignette("visstat\_coreistics"). Throughout, data of class numeric or integer are referred to as numeric, while data of class factor are referred to as categorical. The significance level  $\alpha$  is defined as one minus the confidence level, given by the argument `conf.level`. Assumptions of normality and homoscedasticity are considered met when the corresponding test yields a p-value greater than  $\alpha = 1 - \text{conf.level}$ . The choice of statistical tests performed by `visstat_core()` depends on whether the data are numeric or categorical, the number of levels in the categorical variable, the distribution of the data, and the chosen `conf.level`. The function prioritises interpretable visual output and tests that remain valid under their assumptions, following the logic below:

(1) When the response is numerical and the predictor is categorical, tests of central tendency are performed. If the predictor has two levels: `t.test()` is used if both groups have more than 30

observations (Lumley et al. (2002) <doi:10.1146/annurev.publhealth.23.100901.140546>). For smaller samples, normality is assessed using `shapiro.test()`. If both groups return p-values greater than alpha, `t.test()` is applied; otherwise, `wilcox.test()` is used. For predictors with more than two levels, `aov()` is initially fitted. Residual normality is tested with `shapiro.test()` and `ad.test()`. If  $p > \alpha$  for either test, normality is assumed. Homogeneity of variance is tested with `bartlett.test()`. If  $p > \alpha$ , `aov()` with `TukeyHSD()` is used. If  $p \leq \alpha$ , `oneway.test()` is applied with `TukeyHSD()`. If residuals are not normal, `kruskal.test()` with `pairwise.wilcox.test()` is used.

(2): When both the response and predictor are numerical, a linear model `lm()` is fitted, with residual diagnostics and a confidence band plot.

(3): When both variables are categorical, `visstat_core()` uses `chisq.test()` or `fisher.test()` depending on expected counts, following Cochran's rule (Cochran (1954) <doi:10.2307/3001666>).

Implemented main tests:

`t.test()`, `wilcox.test()`, `aov()`, `oneway.test()`, `lm()`, `kruskal.test()`, `fisher.test()`, `chisq.test()`.

Implemented tests for assumptions:

- Normality: `shapiro.test()` and `ad.test()`
- Heteroscedasticity: `bartlett.test()`

Implemented post hoc tests:

- `TukeyHSD()` for `aov()` and `oneway.test()`
- `pairwise.wilcox.test()` for `kruskal.test()`

## Value

list containing statistics of automatically selected test meeting assumptions. All values are returned as invisible copies. Values can be accessed by assigning a return value to `visstat_core`.

## See Also

See also the package's vignette `vignette("visStatistics")` for the overview, and the accompanying webpage <https://shhschilling.github.io/visStatistics/>.

## Examples

```
# Welch Two Sample t-test (t.test())
visstat_core(mtcars, "mpg", "am")

## Wilcoxon rank sum test (wilcox.test())
grades_gender <- data.frame(
  Sex = as.factor(c(rep("Girl", 20), rep("Boy", 20))),
  Grade = c(
    19.3, 18.1, 15.2, 18.3, 7.9, 6.2, 19.4,
    20.3, 9.3, 11.3, 18.2, 17.5, 10.2, 20.1, 13.3, 17.2, 15.1, 16.2, 17.3,
    16.5, 5.1, 15.3, 17.1, 14.8, 15.4, 14.4, 7.5, 15.5, 6.0, 17.4,
    7.3, 14.3, 13.5, 8.0, 19.5, 13.4, 17.9, 17.7, 16.4, 15.6
  )
)
```

```

)
visstat_core(grades_gender, "Grade", "Sex")

## Welch's oneway ANOVA not assuming equal variances (oneway.test())
anova_npk <- visstat_core(npk, "yield", "block")
anova_npk # prints summary of tests

## Kruskal-Wallis rank sum test (kruskal.test())
visstat_core(iris, "Petal.Width", "Species")
visstat_core(InsectSprays, "count", "spray")

## Linear regression (lm())
visstat_core(trees, "Girth", "Height", conf.level = 0.99)

## Pearson's Chi-squared test (chisq.test())
### Transform array to data.frame
HairEyeColorDataFrame <- counts_to_cases(as.data.frame(HairEyeColor))
visstat_core(HairEyeColorDataFrame, "Hair", "Eye")

## Fisher's exact test (fisher.test())
HairEyeColorMaleFisher <- HairEyeColor[, , 1]
### slicing out a 2 x2 contingency table
blackBrownHazelGreen <- HairEyeColorMaleFisher[1:2, 3:4]
blackBrownHazelGreen <- counts_to_cases(as.data.frame(blackBrownHazelGreen))
fisher_stats <- visstat_core(blackBrownHazelGreen, "Hair", "Eye")
fisher_stats # print out summary statistics

## Saving the graphical output in directory "plotDirectory"
## A) Saving graphical output of type "png" in temporary directory tempdir()
## with default naming convention:
visstat_core(blackBrownHazelGreen, "Hair", "Eye",
  graphicsoutput = "png",
  plotDirectory = tempdir()
)

## Remove graphical output from plotDirectory
file.remove(file.path(tempdir(), "chi_squared_or_fisher_Hair_Eye.png"))
file.remove(file.path(tempdir(), "mosaic_complete_Hair_Eye.png"))

## B) Specifying pdf as output type:
visstat_core(iris, "Petal.Width", "Species",
  graphicsoutput = "pdf",
  plotDirectory = tempdir()
)

## Remove graphical output from plotDirectory
file.remove(file.path(tempdir(), "kruskal_Petal_Width_Species.pdf"))

## C) Specifying "plotName" overwrites default naming convention
visstat_core(iris, "Petal.Width", "Species",
  graphicsoutput = "pdf",
  plotName = "kruskal_iris", plotDirectory = tempdir()
)

```

```
## Remove graphical output from plotDirectory
file.remove(file.path(tempdir(), "kruskal_iris.pdf"))
```

---

vis\_anova\_assumptions *Visualisation of the normality distribution of the standardised residuals of the ANOVA*

---

## Description

vis\_anova\_assumptions checks for normality of the standardised residuals of the ANOVA. Both the Shapiro-Wilk test `shapiro.test()` and the Anderson-Darling test `ad.test()` check the null that the standardised residuals are normally distributed. It generates a scatter plot of the standardised residuals versus the fitted mean values of the linear models for each level of fact. Furthermore a normal QQ plot of the standardised residuals is generated. The null of homogeneity of variances of each factor level is tested with the `bartlett.test()`.

## Usage

```
vis_anova_assumptions(
  samples,
  fact,
  conf.level = 0.95,
  samplename = "",
  factorname = "",
  cex = 1
)
```

## Arguments

samples	vector containing dependent variable, datatype numeric
fact	vector containing independent variable, datatype factor
conf.level	confidence level, 0.95=default
samplename	name of sample used in graphical output, datatype character, ""=default
factorname	name of sample used in graphical output, datatype character, ""=default
cex	number indicating the amount by which plotting text and symbols should be scaled relative to the default. 1=default, 1.5 is 50% larger, 0.5 is 50% smaller, etc.

## Value

list containing the test statistics of the anova, the p values generated by the Shapiro-Wilk test `shapiro.test()`, the Anderson-Darling test `ad.test()` and the `bartlett.test()`.

**Examples**

```
ToothGrowth$dose <- as.factor(ToothGrowth$dose)
vis_anova_assumptions(ToothGrowth$len, ToothGrowth$dose)

vis_anova_assumptions(ToothGrowth$len, ToothGrowth$supp)
vis_anova_assumptions(iris$Petal.Width, iris$Species)
```

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