Package 'jmv'

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Type Package

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Description A suite of common statistical methods such as descriptives, t-tests, ANOVAs, regression, correlation matrices, proportion tests, contingency tables, and factor analysis. This package is also useable from the 'jamovi' statistical spreadsheet (see https://www.jamovi.org for more information).			
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ancova

ANCOVA

Description

The Analysis of Covariance (ANCOVA) is used to explore the relationship between a continuous dependent variable, one or more categorical explanatory variables, and one or more continuous explanatory variables (or covariates). It is essentially the same analysis as ANOVA, but with the addition of covariates.

Usage

```
ancova(data, dep, factors = NULL, covs = NULL, effectSize = NULL,
modelTerms = NULL, ss = "3", homo = FALSE, qq = FALSE,
contrasts = NULL, postHoc = NULL, postHocCorr = list("tukey"),
emMeans = list(list()), emmPlots = TRUE, emmPlotData = FALSE,
emmPlotError = "ci", emmTables = FALSE, emmWeights = TRUE,
ciWidthEmm = 95, formula)
```

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Arguments

data the data as a data frame

dep the dependent variable from data, variable must be numeric (not necessary

when providing a formula, see examples)

factors the explanatory factors in data (not necessary when providing a formula, see

examples)

covs the explanatory covariates (not necessary when providing a formula, see exam-

ples)

effectSize one or more of 'eta', 'partEta', or 'omega'; use eta², partial eta², and omega²

effect sizes, respectively

modelTerms a formula describing the terms to go into the model (not necessary when provid-

ing a formula, see examples)

ss '1', '2' or '3' (default), the sum of squares to use
homo TRUE or FALSE (default), perform homogeneity tests

TRUE or FALSE (default), provide a Q-Q plot of residuals

contrasts a list of lists specifying the factor and type of contrast to use, one of 'deviation',

'simple', 'difference', 'helmert', 'repeated' or 'polynomial'

postHoc a formula containing the terms to perform post-hoc tests on (see the examples) one or more of 'none', 'tukey', 'scheffe', 'bonf', or 'holm'; provide no,

Tukey, Scheffe, Bonferroni, and Holm Post Hoc corrections respectively

emMeans a formula containing the terms to estimate marginal means for (see the exam-

ples)

emmPlots TRUE (default) or FALSE, provide estimated marginal means plots emmPlotData TRUE or FALSE (default), plot the data on top of the marginal means

emmPlotError 'none', 'ci' (default), or 'se'. Use no error bars, use confidence intervals, or

use standard errors on the marginal mean plots, respectively

emmTables TRUE or FALSE (default), provide estimated marginal means tables

emmWeights TRUE (default) or FALSE, weigh each cell equally or weigh them according to the

cell frequency

ciWidthEmm a number between 50 and 99.9 (default: 95) specifying the confidence interval

width for the estimated marginal means

formula (optional) the formula to use, see the examples

Value

A results object containing:

results\$main a table of ANCOVA results results\$model The underlying aov object results\$assump\$homo a table of homogeneity tests

results\$assump\$qq a q-q plot

results\$contrasts an array of contrasts tables results\$postHoc an array of post-hoc tables

results\$emm an array of the estimated marginal means plots + tables

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```
Tables can be converted to data frames with asDF or as.data.frame. For example: results$main$asDF as.data.frame(results$main)
```

Examples

```
data('ToothGrowth')
ancova(formula = len ~ supp + dose, data = ToothGrowth)
  ANCOVA
#
#
 ANCOVA
#
             Sum of Squares df Mean Square F p
#
#
  ______
                     205 1
                                   205.4 11.4 0.001
#
  supp
   dose 2224 1 2224.3 124.0 < .001 Residuals 1023 57 17.9
#
#
#
#
ancova(
   formula = len ~ supp + dose,
   data = ToothGrowth,
   postHoc = \sim supp,
   emMeans = \sim supp)
```

ANOVA ANOVA

Description

The Analysis of Variance (ANOVA) is used to explore the relationship between a continuous dependent variable, and one or more categorical explanatory variables.

Usage

```
ANOVA(data, dep, factors = NULL, effectSize = NULL,
  modelTerms = NULL, ss = "3", homo = FALSE, qq = FALSE,
  contrasts = NULL, postHoc = NULL, postHocCorr = list("tukey"),
  emMeans = list(list()), emmPlots = TRUE, emmPlotData = FALSE,
  emmPlotError = "ci", emmTables = FALSE, emmWeights = TRUE,
  ciWidthEmm = 95, formula)
anova(...)
```

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Arguments

data	the data as a data frame
dep	the dependent variable from data, variable must be numeric (not necessary when providing a formula, see examples)
factors	the explanatory factors in data (not necessary when providing a formula, see examples)
effectSize	one or more of 'eta', 'partEta', or 'omega'; use eta², partial eta², and omega² effect sizes, respectively
modelTerms	a formula describing the terms to go into the model (not necessary when providing a formula, see examples)
SS	'1', '2' or '3' (default), the sum of squares to use
homo	TRUE or FALSE (default), perform homogeneity tests
qq	TRUE or FALSE (default), provide a Q-Q plot of residuals
contrasts	a list of lists specifying the factor and type of contrast to use, one of 'deviation', 'simple', 'difference', 'helmert', 'repeated' or 'polynomial'
postHoc	a formula containing the terms to perform post-hoc tests on (see the examples)
postHocCorr	one or more of 'none', 'tukey', 'scheffe', 'bonf', or 'holm'; provide no, Tukey, Scheffe, Bonferroni, and Holm Post Hoc corrections respectively
emMeans	a formula containing the terms to estimate marginal means for (see the examples)
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmPlotData	TRUE or FALSE (default), plot the data on top of the marginal means
emmPlotError	'none', 'ci' (default), or 'se'. Use no error bars, use confidence intervals, or use standard errors on the marginal mean plots, respectively
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means
formula	(optional) the formula to use, see the examples
	arguments passed to anova() are passed on to ANOVA()

Details

ANOVA assumes that the residuals are normally distributed, and that the variances of all groups are equal. If one is unwilling to assume that the variances are equal, then a Welch's test can be used instead (However, the Welch's test does not support more than one explanatory factor). Alternatively, if one is unwilling to assume that the data is normally distributed, a non-parametric approach (such as Kruskal-Wallis) can be used.

Value

A results object containing:

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```
results$main a table of ANOVA results
results$model The underlying aov object
results$assump$homo a table of homogeneity tests
results$assump$qq a q-q plot
results$contrasts an array of contrasts tables
results$postHoc an array of post-hoc tables
results$emm array of the estimated marginal means plots + tables
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$main\$asDF as.data.frame(results\$main)

```
Examples
```

```
data('ToothGrowth')
ANOVA(formula = len ~ dose * supp, data = ToothGrowth)
   ANOVA
#
   ANOVA
#
#
                      Sum of Squares df Mean Square F p
#
#
     dose

    dose
    2426
    2
    1213.2
    92.00
    < .001</td>

    supp
    205
    1
    205.4
    15.57
    < .001</td>

    dose: supp
    108
    2
    54.2
    4.11
    0.022

    Residuals
    712
    54
    13.2

#
#
#
ANOVA(
     formula = len ~ dose * supp,
     data = ToothGrowth,
     emMeans = \sim supp + dose:supp, # est. marginal means for supp and dose:supp
     emmPlots = TRUE,  # produce plots of those marginal means
     emmTables = TRUE)
                                           # produce tables of those marginal means
```

anovaNP

One-Way ANOVA (Non-parametric)

Description

The Kruskal-Wallis test is used to explore the relationship between a continuous dependent variable, and a categorical explanatory variable. It is analagous to ANOVA, but with the advantage of being non-parametric and having fewer assumptions. However, it has the limitation that it can only test a single explanatory variable at a time.

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Usage

```
anovaNP(data, deps, group, pairs = FALSE, formula)
```

Arguments

data the data as a data frame

deps a string naming the dependent variable in data

group a string naming the grouping or independent variable in data

pairs TRUE or FALSE (default), perform pairwise comparisons

formula (optional) the formula to use, see the examples

Value

A results object containing:

results\$table results\$comparisons

a table of the test results an array of pairwise comparison tables

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$table\$asDF

```
as.data.frame(results$table)
```

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Description

The Analysis of Variance (ANOVA) is used to explore the relationship between a continuous dependent variable, and one or more categorical explanatory variables. This 'One-Way ANOVA' is a simplified version of the 'normal' ANOVA, allowing only a single explanatory factor, however also providing a Welch's ANOVA. The Welch's ANOVA has the advantage that it need not assume that the variances of all groups are equal.

Usage

```
anovaOneW(data, deps, group, welchs = TRUE, fishers = FALSE,
  miss = "perAnalysis", desc = FALSE, descPlot = FALSE,
  norm = FALSE, qq = FALSE, eqv = FALSE, phMethod = "none",
  phMeanDif = TRUE, phSig = TRUE, phTest = FALSE, phFlag = FALSE,
  formula)
```

data	the data as a data frame
deps	a string naming the dependent variables in data
group	a string naming the grouping or independent variable in data
welchs	TRUE (default) or FALSE, perform Welch's one-way ANOVA which does not assume equal variances
fishers	TRUE or FALSE (default), perform Fisher's one-way ANOVA which assumes equal variances
miss	'perAnalysis' or 'listwise', how to handle missing values; 'perAnalysis' excludes missing values for individual dependent variables, 'listwise' excludes a row from all analyses if one of its entries is missing.
desc	TRUE or FALSE (default), provide descriptive statistics
descPlot	TRUE or FALSE (default), provide descriptive plots
norm	TRUE or FALSE (default), perform Shapiro-Wilk test of normality
qq	TRUE or FALSE (default), provide a Q-Q plot of residuals
eqv	TRUE or FALSE (default), perform Levene's test for equality of variances
phMethod	'none', 'gamesHowell' or 'tukey', which post-hoc tests to provide; 'none' shows no post-hoc tests, 'gamesHowell' shows Games-Howell post-hoc tests where no equivalence of variances is assumed, and 'tukey' shows Tukey post-hoc tests where equivalence of variances is assumed
phMeanDif	TRUE (default) or FALSE, provide mean differences for post-hoc tests
phSig	TRUE (default) or FALSE, provide significance levels for post-hoc tests
phTest	TRUE or FALSE (default), provide test results (t-value and degrees of freedom) for post-hoc tests
phFlag	TRUE or FALSE (default), flag significant post-hoc comparisons
formula	(optional) the formula to use, see the examples

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Details

For convenience, this method allows specifying multiple dependent variables, resulting in multiple independent tests.

Note that the Welch's ANOVA is the same procedure as the Welch's independent samples t-test.

Value

A results object containing:

```
results$anova
results$desc
results$assump$norm
results$assump$eqv
results$plots
results$postHoc
```

a table of the test results a table containing the group descriptives a table containing the normality tests a table of equality of variances tests an array of groups of plots an array of post-hoc tables

Tables can be converted to data frames with asDF or as.data.frame. For example:

```
results$anova$asDF
as.data.frame(results$anova)
```

```
data('ToothGrowth')
dat <- ToothGrowth
dat$dose <- factor(dat$dose)</pre>
anovaOneW(formula = len ~ dose, data = dat)
  ONE-WAY ANOVA
#
#
#
  One-Way ANOVA (Welch's)
#
#
               df1 df2 p
#
  -----
#
  len 68.4 2 37.7 < .001
#
```

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Description

The Repeated Measures ANOVA is used to explore the relationship between a continuous dependent variable and one or more categorical explanatory variables, where one or more of the explanatory variables are 'within subjects' (where multiple measurements are from the same subject). Additionally, this analysis allows the inclusion of covariates, allowing for repeated measures ANCOVAs as well.

Usage

```
anovaRM(data, rm = list(list(label = "RM Factor 1", levels =
   list("Level 1", "Level 2"))), rmCells = NULL, bs = NULL,
   cov = NULL, effectSize = NULL, depLabel = "Dependent",
   rmTerms = NULL, bsTerms = NULL, ss = "3", spherTests = FALSE,
   spherCorr = list("none"), leveneTest = FALSE, contrasts = NULL,
   postHoc = NULL, postHocCorr = list("tukey"),
   emMeans = list(list()), emmPlots = TRUE, emmTables = FALSE,
   emmWeights = TRUE, ciWidthEmm = 95, emmPlotData = FALSE,
   emmPlotError = "ci")
```

data	the data as a data frame
rm	a list of lists, where each list describes the label (as a string) and the levels (as vector of strings) of a particular repeated measures factor
rmCells	a list of lists, where each list decribes a repeated measure (as a string) from data defined as measure and the particular combination of levels from rm that it belongs to (as a vector of strings) defined as $cell$
bs	a vector of strings naming the between subjects factors from data
cov	a vector of strings naming the covariates from data. Variables must be numeric
effectSize	one or more of 'eta', 'partEta', or 'omega'; use eta², partial eta², and omega² effect sizes, respectively
depLabel	a string (default: 'Dependent') describing the label used for the dependent variable throughout the analysis
rmTerms	a list of character vectors describing the repeated measures terms to go into the model
bsTerms	a list of character vectors describing the between subjects terms to go into the model
ss	'2' or '3' (default), the sum of squares to use
spherTests	TRUE or FALSE (default), perform sphericity tests
spherCorr	one or more of 'none' (default), 'GG', or HF; use no p-value correction, the Greenhouse-Geisser p-value correction, and the Huynh-Feldt p-value correction for shericity, respectively
leveneTest	TRUE or FALSE (default), test for equality of variances (i.e., Levene's test)
contrasts	in development

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postHoc	a list of character vectors describing the post-hoc tests that need to be computed
postHocCorr	one or more of 'none', 'tukey' (default), 'scheffe', 'bonf', or 'holm'; use no, Tukey, Scheffe, Bonferroni and Holm posthoc corrections, respectively
emMeans	a list of lists specifying the variables for which the estimated marginal means need to be calculate. Supports up to three variables per term.
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency $$
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means $\frac{1}{2}$
emmPlotData	TRUE or FALSE (default), plot the data on top of the marginal means
emmPlotError	'none', 'ci' (default), or 'se'. Use no error bars, use confidence intervals, or use standard errors on the marginal mean plots, respectively

Details

This analysis requires that the data be in 'wide format', where each row represents a subject (as opposed to long format, where each measurement of the dependent variable is represented as a row).

A non-parametric equivalent of the repeated measures ANOVA also exists; the Friedman test. However, it has the limitation of only being able to test a single factor.

Value

A results object containing:

```
results$rmTable a table
results$bsTable a table
results$assump$spherTable a table
results$assump$leveneTable a table
```

results\$contrasts an array of tables results\$postHoc an array of tables

results\$emm an array of the estimated marginal means plots + tables

```
Tables can be converted to data frames with asDF or as.data.frame. For example: results$rmTable$asDF as.data.frame(results$rmTable)
```

```
## Not run:
data('bugs', package = 'jmv')
```

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```
anovaRM(
  data = bugs,
  rm = list(
    list(
       label = 'Frightening',
       levels = c('Low', 'High'))),
  rmCells = list(
     list(
       measure = 'LDLF',
       cell = 'Low'),
     list(
       measure = 'LDHF',
       cell = 'High')),
  rmTerms = list(
     'Frightening'))
#
 REPEATED MEASURES ANOVA
#
#
#
 Within Subjects Effects
 ______
           Sum of Squares df Mean Square F p
#
#
  ______
  Frightening 126 1 126.11 44.2 < .001 Residual 257 90 2.85
#
#
#
   Note. Type 3 Sums of Squares
#
#
#
#
 Between Subjects Effects
#
 ______
      Sum of Squares df Mean Square F p
#
 ______
           954 90 10.6
#
#
   Note. Type 3 Sums of Squares
#
## End(Not run)
```

anovaRMNP

Repeated Measures ANOVA (Non-parametric)

Description

The Friedman test is used to explore the relationship between a continuous dependent variable and a categorical explanatory variable, where the explanatory variable is 'within subjects' (where multiple measurements are from the same subject). It is analagous to Repeated Measures ANOVA, but with the advantage of being non-parametric, and not requiring the assumptions of normality or

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homogeneity of variances. However, it has the limitation that it can only test a single explanatory variable at a time.

Usage

```
anovaRMNP(data, measures, pairs = FALSE, desc = FALSE, plots = FALSE,
plotType = "means")
```

Arguments

data	the data as a data frame
measures	a vector of strings naming the repeated measures variables
pairs	TRUE or FALSE (default), perform pairwise comparisons
desc	TRUE or FALSE (default), provide descriptive statistics
plots	TRUE or FALSE (default), provide a descriptive plot
plotType	'means' (default) or 'medians', the error bars to use in the plot

Value

A results object containing:

results\$table	a table of the Friedman test results
results\$comp	a table of the pairwise comparisons
results\$desc	a table containing the descriptives
results\$plot	a descriptives plot

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$table\$asDF as.data.frame(results\$table)

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bugs data sets

Description

data sets

Author(s)

Ryan, Wilde & Crist (2013)

References

http://faculty.kutztown.edu/rryan/RESEARCH/PUBS/Ryan,%20Wilde,%20%26%20Crist%202013%20Web%20exp%20vs%20lab.pdf

cfa

Confirmatory Factor Analysis

Description

Confirmatory Factor Analysis

Usage

```
cfa(data, factors = list(list(label = "Factor 1", vars = list())),
  resCov, miss = "fiml", constrain = "facVar", estTest = TRUE,
  ci = FALSE, ciWidth = 95, stdEst = FALSE, factCovEst = TRUE,
  factInterceptEst = FALSE, resCovEst = FALSE,
  resInterceptEst = FALSE, fitMeasures = list("cfi", "tli", "rmsea"),
  modelTest = TRUE, pathDiagram = FALSE, corRes = FALSE,
  hlCorRes = 0.1, mi = FALSE, hlMI = 3)
```

data	the data as a data frame
factors	a list containing named lists that define the label of the factor and the vars that belong to that factor
resCov	a list of lists specifying the residual covariances that need to be estimated
miss	'listwise' or 'fiml', how to handle missing values; 'listwise' excludes a row from all analyses if one of its entries is missing, 'fiml' uses a full information maximum likelihood method to estimate the model.
constrain	'facVar' or 'facInd', how to contrain the model; 'facVar' fixes the factor variances to one, 'facInd' fixes each factor to the scale of its first indicator.

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TRUE (default) or FALSE, provide 'Z' and 'p' values for the model estimates estTest TRUE or FALSE (default), provide a confidence interval for the model estimates ci ciWidth a number between 50 and 99.9 (default: 95) specifying the confidence interval width that is used as 'ci' stdEst TRUE or FALSE (default), provide a standardized estimate for the model estimates factCovEst TRUE (default) or FALSE, provide estimates for the factor (co)variances

factInterceptEst

TRUE or FALSE (default), provide estimates for the factor intercepts

resCovEst TRUE (default) or FALSE, provide estimates for the residual (co)variances

resInterceptEst

TRUE or FALSE (default), provide estimates for the residual intercepts

one or more of 'cfi', 'tli', 'srmr', 'rmsea', 'aic', or 'bic'; use CFI, TLI, fitMeasures

SRMR, RMSEA + 90% confidence interval, adjusted AIC, and BIC model fit

measures, respectively

modelTest TRUE (default) or FALSE, provide a chi-square test for exact fit that compares the

model with the perfect fitting model

TRUE or FALSE (default), provide a path diagram of the model pathDiagram

corRes TRUE or FALSE (default), provide the residuals for the observed correlation ma-

trix (i.e., the difference between the expected correlation matrix and the ob-

served correlation matrix)

hlCorRes a number (default: 0.1), highlight values in the 'corRes' table above this value

TRUE or FALSE (default), provide modification indices for the parameters not тi

included in the model

hlMI a number (default: 3), highlight values in the 'modIndices' tables above this

value

Value

A results object containing:

results\$factorLoadings a table containing the factor loadings a table containing factor covariances estimates results\$factorEst\$factorCov results\$factorEst\$factorIntercept a table containing factor intercept estimates results\$resEst\$resCov a table containing residual covariances estimates a table containing residual intercept estimates results\$resEst\$resIntercept

results\$modelFit\$test a table containing the chi-square test for exact fit results\$modelFit\$fitMeasures a table containing fit measures

results\$modelPerformance\$corRes a table containing residuals for the observed correlation matrix

results\$modelPerformance\$modIndices a group

results\$pathDiagram an image containing the model path diagram the lavaan syntax used to fit the model results\$modelSyntax

Tables can be converted to data frames with asDF or as.data.frame. For example:

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```
results$factorLoadings$asDF
as.data.frame(results$factorLoadings)
```

```
data <- lavaan::HolzingerSwineford1939</pre>
jmv::cfa(
   data = data,
   factors = list(
      list(label="Visual", vars=c("x1", "x2", "x3")),
      list(label="Textual", vars=c("x4", "x5", "x6")),
      list(label="Speed", vars=c("x7", "x8", "x9"))),
   resCov = NULL)
#
  CONFIRMATORY FACTOR ANALYSIS
# Factor Loadings
           Indicator Estimate SE
   Factor
                        0.900 0.0832 10.81 < .001
#
   Visual x1
                    0.498 0.0808 6.16 < .001
#
     x2
          x3
                       0.656 0.0776 8.46 < .001
  Textual x4
#
                       0.990 0.0567 17.46 < .001
          x5
                       1.102 0.0626 17.60 < .001
                       0.917 0.0538 17.05 < .001
          x6
  Speed x7
                       0.619 0.0743 8.34 < .001
                        0.731 0.0755 9.68 < .001
           x8
                       0.670 0.0775 8.64 < .001
           x9
#
#
  FACTOR ESTIMATES
#
#
#
  Factor Covariances
#
                    Estimate SE Z p
#
 -----
  Visual Visual
                    1.000 a
           Textual 0.459 0.0635 7.22 < .001
Speed 0.471 0.0862 5.46 < .001
#
#
  Textual Textual 1.000 a
   Speed 0.283
Speed 5peed 1.000 a
                             0.0715 3.96 < .001
#
#
#
#
   a fixed parameter
#
# MODEL FIT
```

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```
# Test for Exact Fit
#
 X<sup>2</sup> df p
# -----
#
 85.3 24 < .001
 -----
# Fit Measures
#
 CFI TLI RMSEA Lower Upper
#
# -----
 0.931 0.896 0.0921 0.0714 0.114
#
#
#
```

contTables

Contingency Tables

Description

X2 test of association

Usage

```
contTables(data, rows, cols, counts = NULL, layers = NULL,
  chiSq = TRUE, chiSqCorr = FALSE, likeRat = FALSE, fisher = FALSE,
  contCoef = FALSE, phiCra = FALSE, logOdds = FALSE, odds = FALSE,
  relRisk = FALSE, ci = TRUE, ciWidth = 95, gamma = FALSE,
  taub = FALSE, exp = FALSE, pcRow = FALSE, pcCol = FALSE,
  pcTot = FALSE, formula)
```

data	the data as a data frame
rows	the variable to use as the rows in the contingency table (not necessary when providing a formula, see the examples)
cols	the variable to use as the columns in the contingency table (not necessary when providing a formula, see the examples)
counts	the variable to use as the counts in the contingency table (not necessary when providing a formula, see the examples)
layers	the variables to use to split the contingency table (not necessary when providing a formula, see the examples)
chiSq	TRUE (default) or FALSE, provide X ²
chiSqCorr	TRUE or FALSE (default), provide X2 with continuity correction

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likeRat	TRUE or FALSE (default), provide the likelihood ratio
fisher	TRUE or FALSE (default), provide Fisher's exact test
contCoef	TRUE or FALSE (default), provide the contingency coefficient
phiCra	TRUE or FALSE (default), provide Phi and Cramer's V
logOdds	TRUE or FALSE (default), provide the log odds ratio (only available for $2x2$ tables)
odds	TRUE or FALSE (default), provide the odds ratio (only available for 2x2 tables)
relRisk	TRUE or FALSE (default), provide the relative risk (only available for $2x2$ tables)
ci	TRUE or FALSE (default), provide confidence intervals for the comparative measures $% \left(\frac{1}{2}\right) =\left(\frac{1}{2}\right) \left(\frac{1}{$
ciWidth	a number between 50 and 99.9 (default: 95), width of the confidence intervals to provide $$
gamma	TRUE or FALSE (default), provide gamma
taub	TRUE or FALSE (default), provide Kendall's tau-b
exp	TRUE or FALSE (default), provide the expected counts
pcRow	TRUE or FALSE (default), provide row percentages
pcCol	TRUE or FALSE (default), provide column percentages
pcTot	TRUE or FALSE (default), provide total percentages
formula	(optional) the formula to use, see the examples

Value

A results object containing:

```
results$freqs a table of proportions
results$chiSq a table of X² test results
results$nom a table of comparative measures
results$nom a table of the 'nominal' test results
results$gamma a table of the gamma test results
results$taub a table of the Kendall's tau-b test results
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$freqs\$asDF

```
as.data.frame(results$freqs)
```

```
data('HairEyeColor')
dat <- as.data.frame(HairEyeColor)
contTables(formula = Freq ~ Hair:Eye, dat)
#
# CONTINGENCY TABLES</pre>
```

contTablesPaired 19

```
#
   Contingency Tables
#
#
      Hair Brown Blue Hazel Green Total
   _____

    Black
    68
    20
    15
    5
    108

    Brown
    119
    84
    54
    29
    286

    Red
    26
    17
    14
    14
    71

    Blond
    7
    94
    10
    16
    127

    Total
    220
    215
    93
    64
    592

#
   X<sup>2</sup> Tests
#
#
            Value df p
#
     χ²
           138 9 < .001
#
     N
              592
# Alternatively, omit the left of the formula (`Freq`) if each row
# represents a single observation:
contTables(formula = ~ Hair:Eye, dat)
```

contTablesPaired

Paired Samples Contingency Tables

Description

McNemar test

Usage

```
contTablesPaired(data, rows, cols, counts = NULL, chiSq = TRUE,
  chiSqCorr = FALSE, exact = FALSE, pcRow = FALSE, pcCol = FALSE,
  formula)
```

data	the data as a data frame
rows	the variable to use as the rows in the contingency table (not necessary when providing a formula, see the examples)
cols	the variable to use as the columns in the contingency table (not necessary when providing a formula, see the examples)

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counts	the variable to use as the counts in the contingency table (not necessary when providing a formula, see the examples)
chiSq	TRUE (default) or FALSE, provide X ²
chiSqCorr	TRUE or FALSE (default), provide X2 with continuity correction
exact	TRUE or FALSE (default), provide an exact log odds ratio (requires exact2x2 to be installed)
pcRow	TRUE or FALSE (default), provide row percentages
pcCol	TRUE or FALSE (default), provide column percentages
formula	(optional) the formula to use, see the examples

Value

A results object containing:

results\$freqs a proportions table results\$test a table of test results

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$freqs\$asDF

```
as.data.frame(results$freqs)
```

```
dat <- data.frame(</pre>
    `1st survey` = c('Approve', 'Approve', 'Disapprove'),
    '2nd survey' = c('Approve', 'Disapprove', 'Approve', 'Disapprove'),
    `Counts` = c(794, 150, 86, 570),
    check.names=FALSE)
contTablesPaired(formula = Counts ~ `1st survey`: `2nd survey`, data = dat)
#
   PAIRED SAMPLES CONTINGENCY TABLES
#
#
#
  Contingency Tables
   -----
   1st survey Approve Disapprove Total
#
#

        Approve
        794
        150
        944

        Disapprove
        86
        570
        656

        Total
        880
        720
        1600

#
#
#
#
   ______
#
#
# McNemar Test
#
                                 Value df p
#
```

corrMatrix 21

corrMatrix

Correlation Matrix

Description

Correlation matrices are a way to examine linear relationships between two or more continuous variables.

Usage

```
corrMatrix(data, vars, pearson = TRUE, spearman = FALSE,
  kendall = FALSE, sig = TRUE, flag = FALSE, ci = FALSE,
  ciWidth = 95, plots = FALSE, plotDens = FALSE, plotStats = FALSE,
  hypothesis = "corr")
```

data	the data as a data frame
vars	a vector of strings naming the variables to correlate in data
pearson	TRUE (default) or FALSE, provide Pearson's R
spearman	TRUE or FALSE (default), provide Spearman's rho
kendall	TRUE or FALSE (default), provide Kendall's tau-b
sig	TRUE (default) or FALSE, provide significance levels
flag	TRUE or FALSE (default), flag significant correlations
ci	TRUE or FALSE (default), provide confidence intervals
ciWidth	a number between 50 and 99.9 (default: 95), the width of confidence intervals to provide
plots	TRUE or FALSE (default), provide a correlation matrix plot
plotDens	TRUE or FALSE (default), provide densities in the correlation matrix plot
plotStats	TRUE or FALSE (default), provide statistics in the correlation matrix plot
hypothesis	one of 'corr' (default), 'pos', 'neg' specifying the alernative hypothesis; correlated, correlated positively, correlated negatively respectively.

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Details

For each pair of variables, a Pearson's r value indicates the strength and direction of the relationship between those two variables. A positive value indicates a positive relationship (higher values of one variable predict higher values of the other variable). A negative Pearson's r indicates a negative relationship (higher values of one variable predict lower values of the other variable, and viceversa). A value of zero indicates no relationship (whether a variable is high or low, does not tell us anything about the value of the other variable).

More formally, it is possible to test the null hypothesis that the correlation is zero using and calculate a p-value. If the p-value is low, it suggests the correlation co-efficient is not zero, and there is a linear (or more complex) relationship between the two variables.

Value

A results object containing:

```
results$matrix a correlation matrix table results$plot a correlation matrix plot
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$matrix\$asDF

```
as.data.frame(results$matrix)
```

```
## Not run:
data('mtcars')
corrMatrix(mtcars, vars = vars(mpg, cyl, disp, hp))
  CORRELATION MATRIX
#
#
#
  Correlation Matrix
#
                             cyl disp
#
                      mpg
                                              hp
#
  _____
                             -0.852 -0.848
                                           -0.776
#
   mpg
          Pearson's r
                             < .001
#
          p-value
                                     < .001
                                             < .001
#
          Pearson's r
                                      0.902
                                              0.832
#
   cyl
                                     < .001
#
          p-value
                                              < .001
#
#
   disp
          Pearson's r
                                               0.791
          p-value
                                              < .001
#
#
          Pearson's r
    dd
#
          p-value
```

descriptives 23

```
## End(Not run)
```

descriptives Descriptives

Description

Descriptives are an assortment of summarising statistics, and visualizations which allow exploring the shape and distribution of data. It is good practice to explore your data with descriptives before proceeding to more formal tests.

Usage

```
descriptives(data, vars, splitBy = NULL, freq = FALSE, hist = FALSE,
  dens = FALSE, bar = FALSE, barCounts = FALSE, box = FALSE,
  violin = FALSE, dot = FALSE, dotType = "jitter", qq = FALSE,
  n = TRUE, missing = TRUE, mean = TRUE, median = TRUE,
  mode = FALSE, sum = FALSE, sd = FALSE, variance = FALSE,
  range = FALSE, min = TRUE, max = TRUE, se = FALSE,
  skew = FALSE, kurt = FALSE, sw = FALSE, quart = FALSE,
  pcEqGr = FALSE, pcNEqGr = 4, formula)
```

data	the data as a data frame
vars	a vector of strings naming the variables of interest in data
splitBy	a vector of strings naming the variables used to split vars
freq	TRUE or FALSE (default), provide frequency tables (nominal, ordinal variables only)
hist	TRUE or FALSE (default), provide histograms (continuous variables only)
dens	TRUE or FALSE (default), provide density plots (continuous variables only)
bar	TRUE or FALSE (default), provide bar plots (nominal, ordinal variables only)
barCounts	TRUE or FALSE (default), add counts to the bar plots
box	TRUE or FALSE (default), provide box plots (continuous variables only)
violin	TRUE or FALSE (default), provide violin plots (continuous variables only)
dot	TRUE or FALSE (default), provide dot plots (continuous variables only)
dotType	
qq	TRUE or FALSE (default), provide Q-Q plots (continuous variables only)
n	TRUE (default) or FALSE, provide the sample size
missing	TRUE (default) or FALSE, provide the number of missing values
mean	TRUE (default) or FALSE, provide the mean

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median TRUE (default) or FALSE, provide the median mode TRUE or FALSE (default), provide the mode sum TRUE or FALSE (default), provide the sum

sd TRUE or FALSE (default), provide the standard deviation

variance TRUE or FALSE (default), provide the variance
range TRUE or FALSE (default), provide the range
min TRUE or FALSE (default), provide the minimum
max TRUE or FALSE (default), provide the maximum
se TRUE or FALSE (default), provide the standard error
skew TRUE or FALSE (default), provide the skewness
kurt TRUE or FALSE (default), provide the kurtosis

sw TRUE or FALSE (default), provide Shapiro-Wilk p-value

quart TRUE or FALSE (default), provide quartiles pcEqGr TRUE or FALSE (default), provide quantiles

pcNEqGr an integer (default: 4) specifying the number of equal groups

formula (optional) the formula to use, see the examples

Value

A results object containing:

results\$descriptives results\$frequencies results\$plots a table of the descriptive statistics an array of frequency tables an array of descriptive plots

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$descriptives\$asDF as.data.frame(results\$descriptives)

```
## Not run:
data('mtcars')
dat <- mtcars

# frequency tables can be provided for factors
dat$gear <- as.factor(dat$gear)

descriptives(dat, vars = vars(mpg, cyl, disp, gear), freq = TRUE)

# 
# DESCRIPTIVES
#</pre>
```

efa 25

```
Descriptives
#
             mpg cyl disp gear
#
   N 32 32 32 32
Missing 0 0 0 0
#
#

    Mean
    20.1
    6.19
    231
    3.69

    Median
    19.2
    6.00
    196
    4.00

  Minimum 10.4 4.00 71.1
                                       3
  Maximum 33.9 8.00 472 5
#
#
  FREQUENCIES
#
#
  Frequencies of gear
#
  _____
#
    Levels Counts
#
#
                 15
                12
                  5
#
#
# spliting by a variable
descriptives(formula = disp + mpg ~ cyl, dat,
   median=F, min=F, max=F, n=F, missing=F)
# providing histograms
descriptives(formula = mpg ~ cyl, dat, hist=T,
   median=F, min=F, max=F, n=F, missing=F)
# splitting by multiple variables
descriptives(formula = mpg ~ cyl:gear, dat,
   median=F, min=F, max=F, missing=F)
## End(Not run)
```

efa

Exploratory Factor Analysis

Description

Exploratory Factor Analysis

Usage

```
efa(data, vars, nFactorMethod = "parallel", nFactors = 1,
    minEigen = 1, extraction = "minres", rotation = "oblimin",
```

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```
hideLoadings = 0.3, sortLoadings = FALSE, screePlot = FALSE,
eigen = FALSE, factorCor = FALSE, factorSummary = FALSE,
modelFit = FALSE, kmo = FALSE, bartlett = FALSE)
```

Arguments

data the data as a data frame

vars a vector of strings naming the variables of interest in data

nFactorMethod 'parallel' (default), 'eigen' or 'fixed', the way to determine the number

of factors

nFactors an integer (default: 1), the number of factors in the model

minEigen a number (default: 1), the minimal eigenvalue for a factor to be included in the

model

extraction 'minres' (default), 'ml', or 'pa' use respectively 'minimum residual', 'maxi-

mum likelihood', or 'prinicipal axis' as the factor extraction method

rotation 'none', 'varimax', 'quartimax', 'promax', 'oblimin' (default), or 'simplimax',

the rotation to use in estimation

hideLoadings a number (default: 0.3), hide factor loadings below this value

sortLoadings TRUE or FALSE (default), sort the factor loadings by size

 ${\tt screePlot} \qquad {\tt TRUE} \ or \ {\tt FALSE} \ (default), \ show \ scree \ plot$

eigen TRUE or FALSE (default), show eigenvalue table factorCor TRUE or FALSE (default), show factor correlations factorSummary TRUE or FALSE (default), show factor summary

modelFit TRUE or FALSE (default), show model fit measures and test

kmo TRUE or FALSE (default), show Kaiser-Meyer-Olkin (KMO) measure of sampling

adequacy (MSA) results

bartlett TRUE or FALSE (default), show Bartlett's test of sphericity results

Value

A results object containing:

results\$text a preformatted

```
data('iris')
efa(iris, vars = vars(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width))
#
# EXPLORATORY FACTOR ANALYSIS
#
# Factor Loadings
```

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#				
#		1	2	Uniqueness
#				
#	Sepal.Length	0.993		0.10181
#	Sepal.Width		0.725	0.42199
#	Petal.Length	0.933		0.00483
#	Petal.Width	0.897		0.07088
#				
#	Note. 'oblimin'	rotation	was used	
#				

linReg

Linear Regression

Description

Linear Regression

Usage

```
linReg(data, dep, covs = NULL, factors = NULL, blocks = list(list()),
  refLevels = NULL, r = TRUE, r2 = TRUE, r2Adj = FALSE,
  aic = FALSE, bic = FALSE, rmse = FALSE, modelTest = FALSE,
  anova = FALSE, ci = FALSE, ciWidth = 95, stdEst = FALSE,
  ciStdEst = FALSE, ciWidthStdEst = 95, coefPlot = FALSE,
  qqPlot = FALSE, resPlots = FALSE, durbin = FALSE, collin = FALSE,
  cooks = FALSE, emMeans = list(list()), ciEmm = TRUE,
  ciWidthEmm = 95, emmPlots = TRUE, emmTables = FALSE,
  emmWeights = TRUE)
```

data	the data as a data frame
dep	the dependent variable from data, variable must be numeric
covs	the covariates from data
factors	the fixed factors from data
blocks	a list containing vectors of strings that name the predictors that are added to the model. The elements are added to the model according to their order in the list
refLevels	a list of lists specifying reference levels of the dependent variable and all the factors
r	TRUE (default) or FALSE, provide the statistical measure R for the models
r2	TRUE (default) or FALSE, provide the statistical measure R-squared for the models
r2Adj	TRUE or FALSE (default), provide the statistical measure adjusted R-squared for the models \ensuremath{R}

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aic	TRUE or FALSE (default), provide Aikaike's Information Criterion (AIC) for the
	models
bic	TRUE or FALSE (default), provide Bayesian Information Criterion (BIC) for the models
rmse	TRUE or FALSE (default), provide RMSE for the models
modelTest	TRUE (default) or FALSE, provide the model comparison between the models and the NULL model $$
anova	TRUE or FALSE (default), provide the omnibus ANOVA test for the predictors
ci	TRUE or FALSE (default), provide a confidence interval for the model coefficients
ciWidth	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
stdEst	TRUE or FALSE (default), provide a standardized estimate for the model coefficients
ciStdEst	TRUE or FALSE (default), provide a confidence interval for the model coefficient standardized estimates
ciWidthStdEst	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
coefPlot	TRUE or FALSE (default), provide a coefficient plot where for each predictor the estimated coefficient and confidence intervals are plotted.
qqPlot	TRUE or FALSE (default), provide a Q-Q plot of residuals
resPlots	TRUE or FALSE (default), provide residual plots where the dependent variable and each covariate is plotted against the standardized residuals.
durbin	TRUE or FALSE (default), provide results of the Durbin- Watson test for autocorrelation
collin	TRUE or FALSE (default), provide VIF and tolerence collinearity statistics
cooks	TRUE or FALSE (default), provide summary statistics for the Cook's distance
emMeans	a formula containing the terms to estimate marginal means for, supports up to three variables per term
ciEmm	TRUE (default) or FALSE, provide a confidence interval for the estimated marginal means
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency

Value

A results object containing:

results\$modelFit	a table
results\$modelComp	a table

results\$models an array of model specific results

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```
Tables can be converted to data frames with asDF or as.data.frame. For example: results$modelFit$asDF as.data.frame(results$modelFit)
```

Examples

```
data('Prestige', package='carData')
linReg(data = Prestige, dep = income,
         covs = vars(education, prestige, women),
         blocks = list(list('education', 'prestige', 'women')))
#
#
   LINEAR REGRESSION
#
  Model Fit Measures
#
#
   Model R R<sup>2</sup>
#
   _____
      1 0.802 0.643
#
   _____
#
#
#
   MODEL SPECIFIC RESULTS
#
#
   MODEL 1
#
#
#
   Model Coefficients
#
#
     Predictor Estimate SE t
                                                                р
   -----
#

        Intercept
        -253.8
        1086.16
        -0.234
        0.816

        women
        -50.9
        8.56
        -5.948
        < .001</td>

        prestige
        141.4
        29.91
        4.729
        < .001</td>

        education
        177.2
        187.63
        0.944
        0.347

#
#
#
```

logLinear

Log-Linear Regression

Description

Log-Linear Regression

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Usage

```
logLinear(data, factors = NULL, counts = NULL, blocks = list(list()),
  refLevels = NULL, modelTest = FALSE, dev = TRUE, aic = TRUE,
  bic = FALSE, pseudoR2 = list("r2mf"), omni = FALSE, ci = FALSE,
  ciWidth = 95, RR = FALSE, ciRR = FALSE, ciWidthRR = 95,
  emMeans = list(list()), ciEmm = TRUE, ciWidthEmm = 95,
  emmPlots = TRUE, emmTables = FALSE, emmWeights = TRUE)
```

data	the data as a data frame
factors	a vector of strings naming the factors from data
counts	a string naming a variable in data containing counts, or NULL if each row represents a single observation
blocks	a list containing vectors of strings that name the predictors that are added to the model. The elements are added to the model according to their order in the list
refLevels	a list of lists specifying reference levels of the dependent variable and all the factors
modelTest	TRUE or FALSE (default), provide the model comparison between the models and the NULL model $$
dev	TRUE (default) or FALSE, provide the deviance (or -2LogLikelihood) for the models
aic	TRUE (default) or FALSE, provide Aikaike's Information Criterion (AIC) for the models
bic	TRUE or FALSE (default), provide Bayesian Information Criterion (BIC) for the models
pseudoR2	one or more of 'r2mf', 'r2cs', or 'r2n'; use McFadden's, Cox & Snell, and Nagelkerke pseudo-R², respectively
omni	TRUE or FALSE (default), provide the omnibus likelihood ratio tests for the predictors
ci	TRUE or FALSE (default), provide a confidence interval for the model coefficient estimates
ciWidth	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
RR	TRUE or FALSE (default), provide the exponential of the log-rate ratio estimate, or the rate ratio estimate
ciRR	TRUE or FALSE (default), provide a confidence interval for the model coefficient rate ratio estimates
ciWidthRR	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
emMeans	a list of lists specifying the variables for which the estimated marginal means need to be calculate. Supports up to three variables per term.

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ciEmm	TRUE (default) or FALSE, provide a confidence interval for the estimated marginal means
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency

Value

A results object containing:

```
results$modelFit a table
results$modelComp a table
results$models an array of model specific results
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$modelFit\$asDF as.data.frame(results\$modelFit)

```
data('mtcars')
tab <- table('gear'=mtcars$gear, 'cyl'=mtcars$cyl)</pre>
dat <- as.data.frame(tab)</pre>
logLinear(data = dat, factors = vars(gear, cyl), counts = Freq,
        blocks = list(list("gear", "cyl", c("gear", "cyl"))),
        refLevels = list(
            list(var="gear", ref="3"),
            list(var="cyl", ref="4")))
#
 LOG-LINEAR REGRESSION
# Model Fit Measures
#
  Model Deviance AIC R<sup>2</sup>-McF
#
 _____
#
      1 4.12e-10 41.4 1.000
#
  _____
#
#
#
# MODEL SPECIFIC RESULTS
#
# MODEL 1
```

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```
Model Coefficients
#
        Predictor Estimate SE Z
        Intercept -4.71e-16 1.00 -4.71e-16
        gear:

      2.079
      1.06
      1.961

      0.693
      1.22
      0.566

       4 3
                                                                                                            0.050
       5 3
                                                                                                            0.571
        cyl:

      0.693
      1.22

      2.485
      1.04

                                                                                 0.566
2.387
                                                                                                            0.571
       6 4
        8 4
                                                                                         2.387
                                                                                                            0.017
        gear:cyl:

    (4 3):(6 4)
    -1.386
    1.37
    -1.012
    0.311

    (5 3):(6 4)
    -1.386
    1.73
    -0.800
    0.423

    (4 3):(8 4)
    -26.867
    42247.17
    -6.36e -4
    0.999

    (5 3):(8 4)
    -2.485
    1.44
    -1.722
    0.085

#
#
#
```

logRegBin

Binomial Logistic Regression

Description

Binomial Logistic Regression

Usage

```
logRegBin(data, dep, covs = NULL, factors = NULL,
blocks = list(list()), refLevels = NULL, modelTest = FALSE,
dev = TRUE, aic = TRUE, bic = FALSE, pseudoR2 = list("r2mf"),
omni = FALSE, ci = FALSE, ciWidth = 95, OR = FALSE,
ciOR = FALSE, ciWidthOR = 95, emMeans = list(list()),
ciEmm = TRUE, ciWidthEmm = 95, emmPlots = TRUE,
emmTables = FALSE, emmWeights = TRUE, class = FALSE, acc = FALSE,
spec = FALSE, sens = FALSE, auc = FALSE, rocPlot = FALSE,
cutOff = 0.5, cutOffPlot = FALSE, collin = FALSE,
boxTidwell = FALSE, cooks = FALSE)
```

data	the data as a data frame
dep	a string naming the dependent variable from data, variable must be a factor
covs	a vector of strings naming the covariates from data
factors	a vector of strings naming the fixed factors from data

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blocks	a list containing vectors of strings that name the predictors that are added to the model. The elements are added to the model according to their order in the list
refLevels	a list of lists specifying reference levels of the dependent variable and all the factors
modelTest	TRUE or FALSE (default), provide the model comparison between the models and the NULL model
dev	TRUE (default) or FALSE, provide the deviance (or -2LogLikelihood) for the models
aic	TRUE (default) or FALSE, provide Aikaike's Information Criterion (AIC) for the models
bic	TRUE or FALSE (default), provide Bayesian Information Criterion (BIC) for the models
pseudoR2	one or more of 'r2mf', 'r2cs', or 'r2n'; use McFadden's, Cox & Snell, and Nagelkerke pseudo-R², respectively
omni	TRUE or FALSE (default), provide the omnibus likelihood ratio tests for the predictors
ci	TRUE or FALSE (default), provide a confidence interval for the model coefficient estimates
ciWidth	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
OR	TRUE or FALSE (default), provide the exponential of the log-odds ratio estimate, or the odds ratio estimate
ciOR	TRUE or FALSE (default), provide a confidence interval for the model coefficient odds ratio estimates
ciWidthOR	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
emMeans	a list of lists specifying the variables for which the estimated marginal means need to be calculate. Supports up to three variables per term.
ciEmm	TRUE (default) or FALSE, provide a confidence interval for the estimated marginal means
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency
class	TRUE or FALSE (default), provide a predicted classification table (or confusion matrix)
acc	TRUE or FALSE (default), provide the predicted accuracy of outcomes grouped by the cut-off value
spec	TRUE or FALSE (default), provide the predicted specificity of outcomes grouped by the cut-off value

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sens	TRUE or FALSE (default), provide the predicted sensitivity of outcomes grouped by the cut-off value
auc	TRUE or FALSE (default), provide the rea under the ROC curve (AUC)
rocPlot	TRUE or FALSE (default), provide a ROC curve plot
cut0ff	TRUE or FALSE (default), set a cut-off used for the predictions
cutOffPlot	TRUE or FALSE (default), provide a cut-off plot
collin	TRUE or FALSE (default), provide VIF and tolerence collinearity statistics
boxTidwell	TRUE or FALSE (default), provide Box-Tidwell test for linearity of the logit
cooks	TRUE or FALSE (default), provide summary statistics for the Cook's distance

Value

A results object containing:

```
results$modelFit a table
results$modelComp a table
results$models an array of model specific results
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$modelFit\$asDF as.data.frame(results\$modelFit)

```
data('birthwt', package='MASS')
dat <- data.frame(</pre>
           low = factor(birthwt$low),
           age = birthwt$age,
           bwt = birthwt$bwt)
logRegBin(data = dat, dep = low,
         covs = vars(age, bwt),
         blocks = list(list("age", "bwt")),
         refLevels = list(list(var="low", ref="0")))
  BINOMIAL LOGISTIC REGRESSION
#
#
#
 Model Fit Measures
    Model Deviance AIC R2-McF
#
       1 4.97e-7 6.00 1.000
#
#
#
```

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```
# MODEL SPECIFIC RESULTS
#
# MODEL 1
#
# Model Coefficients
    Predictor Estimate SE Z p
  Intercept 2974.73225 218237.2 0.0136 0.989
#
   age -0.00653 482.7 -1.35e-5 1.000
bwt -1.18532 87.0 -0.0136 0.989
#
#
#
    Note. Estimates represent the log odds of "low = 1"
    vs. "low = 0"
#
#
```

logRegMulti

Multinomial Logistic Regression

Description

Multinomial Logistic Regression

Usage

```
logRegMulti(data, dep, covs = NULL, factors = NULL,
  blocks = list(list()), refLevels = NULL, modelTest = FALSE,
  dev = TRUE, aic = TRUE, bic = FALSE, pseudoR2 = list("r2mf"),
  omni = FALSE, ci = FALSE, ciWidth = 95, OR = FALSE,
  ciOR = FALSE, ciWidthOR = 95, emMeans = list(list()),
  ciEmm = TRUE, ciWidthEmm = 95, emmPlots = TRUE,
  emmTables = FALSE, emmWeights = TRUE)
```

data	the data as a data frame
dep	a string naming the dependent variable from data, variable must be a factor
covs	a vector of strings naming the covariates from data
factors	a vector of strings naming the fixed factors from data
blocks	a list containing vectors of strings that name the predictors that are added to the model. The elements are added to the model according to their order in the list
refLevels	a list of lists specifying reference levels of the dependent variable and all the factors
modelTest	TRUE or FALSE (default), provide the model comparison between the models and the NULL model

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dev	TRUE (default) or FALSE, provide the deviance (or -2LogLikelihood) for the models
aic	TRUE (default) or FALSE, provide Aikaike's Information Criterion (AIC) for the models
bic	TRUE or FALSE (default), provide Bayesian Information Criterion (BIC) for the models
pseudoR2	one or more of 'r2mf', 'r2cs', or 'r2n'; use McFadden's, Cox & Snell, and Nagelkerke pseudo-R2, respectively
omni	TRUE or FALSE (default), provide the omnibus likelihood ratio tests for the predictors $% \left(1\right) =\left(1\right) \left(1\right) \left$
ci	TRUE or FALSE (default), provide a confidence interval for the model coefficient estimates
ciWidth	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
OR	TRUE or FALSE (default), provide the exponential of the log-odds ratio estimate, or the odds ratio estimate
ciOR	TRUE or FALSE (default), provide a confidence interval for the model coefficient odds ratio estimates
ciWidthOR	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
emMeans	a list of lists specifying the variables for which the estimated marginal means need to be calculate. Supports up to three variables per term.
ciEmm	TRUE (default) or FALSE, provide a confidence interval for the estimated marginal means $$
ciWidthEmm	a number between 50 and 99.9 (default: 95) specifying the confidence interval width for the estimated marginal means
emmPlots	TRUE (default) or FALSE, provide estimated marginal means plots
emmTables	TRUE or FALSE (default), provide estimated marginal means tables
emmWeights	TRUE (default) or FALSE, weigh each cell equally or weigh them according to the cell frequency $\frac{1}{2}$

Value

A results object containing:

results\$modelFit a table results\$modelComp a table

results\$models an array of model specific results

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$modelFit\$asDF as.data.frame(results\$modelFit)

logRegOrd 37

```
data('birthwt', package='MASS')
dat <- data.frame(</pre>
          race = factor(birthwt$race),
          age = birthwt$age,
          low = factor(birthwt$low))
logRegMulti(data = dat, dep = race,
          covs = age, factors = low,
          blocks = list(list("age", "low")),
          refLevels = list(
              list(var="race", ref="1"),
              list(var="low", ref="0")))
  MULTINOMIAL LOGISTIC REGRESSION
 Model Fit Measures
  Model Deviance AIC R<sup>2</sup>-McF
#
   .....
#
                360 372 0.0333
    1
#
#
  MODEL SPECIFIC RESULTS
  MODEL 1
#
 Model Coefficients
           Predictor Estimate SE Z
   race
 _____
#
   2 - 1 Intercept 0.8155 1.1186 0.729 0.466 age -0.1038 0.0487 -2.131 0.033
#
            low:
#
  1 0 0.7527 0.4700 1.601 0.109
3 - 1 Intercept 1.0123 0.7798 1.298 0.194
age -0.0663 0.0324 -2.047 0.041
#
            low:
            1 0 0.5677 0.3522 1.612 0.107
```

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Description

Ordinal Logistic Regression

Usage

```
logRegOrd(data, dep, covs = NULL, factors = NULL,
blocks = list(list()), refLevels = NULL, modelTest = FALSE,
dev = TRUE, aic = TRUE, bic = FALSE, pseudoR2 = list("r2mf"),
omni = FALSE, thres = FALSE, ci = FALSE, ciWidth = 95,
OR = FALSE, ciOR = FALSE, ciWidthOR = 95)
```

data	the data as a data frame
dep	a string naming the dependent variable from data, variable must be a factor
covs	a vector of strings naming the covariates from data
factors	a vector of strings naming the fixed factors from data
blocks	a list containing vectors of strings that name the predictors that are added to the model. The elements are added to the model according to their order in the list
refLevels	a list of lists specifying reference levels of the dependent variable and all the factors
modelTest	TRUE or FALSE (default), provide the model comparison between the models and the NULL model $$
dev	TRUE (default) or FALSE, provide the deviance (or -2LogLikelihood) for the models
aic	TRUE (default) or FALSE, provide Aikaike's Information Criterion (AIC) for the models $$
bic	TRUE or FALSE (default), provide Bayesian Information Criterion (BIC) for the models $$
pseudoR2	one or more of 'r2mf', 'r2cs', or 'r2n'; use McFadden's, Cox & Snell, and Nagelkerke pseudo-R2, respectively
omni	TRUE or FALSE (default), provide the omnibus likelihood ratio tests for the predictors $% \left(1\right) =\left(1\right) \left(1\right) \left$
thres	TRUE or FALSE (default), provide the thresholds that are used as cut-off scores for the levels of the dependent variable
ci	TRUE or FALSE (default), provide a confidence interval for the model coefficient estimates
ciWidth	a number between 50 and 99.9 (default: 95) specifying the confidence interval width
OR	TRUE or FALSE (default), provide the exponential of the log-odds ratio estimate, or the odds ratio estimate
ciOR	TRUE or FALSE (default), provide a confidence interval for the model coefficient odds ratio estimates
ciWidthOR	a number between 50 and 99.9 (default: 95) specifying the confidence interval width

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Value

A results object containing:

```
results$modelFit a table
results$modelComp a table
results$models an array of model specific results
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$modelFit\$asDF as.data.frame(results\$modelFit)

```
set.seed(1337)
y <- factor(sample(1:3, 100, replace = TRUE))
x1 <- rnorm(100)
x2 <- rnorm(100)
df \leftarrow data.frame(y=y, x1=x1, x2=x2)
logRegOrd(data = df, dep = y,
         covs = vars(x1, x2),
         blocks = list(list("x1", "x2")))
  ORDINAL LOGISTIC REGRESSION
#
# Model Fit Measures
#
   Model Deviance AIC R<sup>2</sup>-McF
#
#
    -----
#
    1 218 226 5.68e-4
#
#
 MODEL SPECIFIC RESULTS
#
  MODEL 1
#
#
 Model Coefficients
#
    Predictor Estimate SE Z p
#
 -----
#

      0.0579
      0.193
      0.300
      0.764

      0.0330
      0.172
      0.192
      0.848

#
#
#
#
#
```

40 mancova

Description

Multivariate Analysis of Covariance

Usage

```
mancova(data, deps, factors = NULL, covs = NULL,
  multivar = list("pillai", "wilks", "hotel", "roy"), boxM = FALSE,
  shapiro = FALSE, qqPlot = FALSE)
```

Arguments

data	the data as a data frame
deps	a string naming the dependent variable from data, variable must be numeric
factors	a vector of strings naming the factors from data
covs	a vector of strings naming the covariates from data
multivar	one or more of 'pillai', 'wilks', 'hotel', or 'roy'; use Pillai's Trace, Wilks' Lambda, Hotelling's Trace, and Roy's Largest Root multivariate statistics, respectively
boxM	TRUE or FALSE (default), provide Box's M test
shapiro	TRUE or FALSE (default), provide Shapiro-Wilk test
qqPlot	TRUE or FALSE (default), provide a Q-Q plot of multivariate normality

Value

A results object containing:

```
results$multivar a table
results$univar a table
results$assump$boxM a table
results$assump$shapiro a table
results$assump$qqPlot an image
```

Tables can be converted to data frames with asDF or as.data.frame. For example:

```
results$multivar$asDF
as.data.frame(results$multivar)
```

```
data('iris')
```

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```
mancova(data = iris,
     deps = vars(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width),
     factors = Species)
   MANCOVA
#
   Multivariate Tests
                                                    value F df1 df2 p
#

      Species
      Pillai's Trace
      1.19
      53.5
      8
      290
      < .001</th>

      Wilks' Lambda
      0.0234
      199
      8
      288
      < .001</td>

      Hotelling's Trace
      32.5
      581
      8
      286
      < .001</td>

      Roy's Largest Root
      32.2
      1167
      4
      145
      < .001</td>

#
#
#
#
#
   Univariate Tests
#
                       Dependent Variable Sum of Squares df Mean Square F
#
      .______
                                                         63.21 2 31.6061 119.3 < .001

11.34 2 5.6725 49.2 < .001

437.10 2 218.5514 1180.2 < .001

80.41 2 40.2067 960.0 < .001

38.96 147 0.2650

16.96 147 0.1154
      Species Sepal.Length
#
                     Sepal.Width
#
                     Petal.Length
#
                      Petal.Width
    Residuals Sepal.Length
                                                                   16.96 147 0.1154
27.22 147 0.1852
6.16 147 0.0419
                         Sepal.Width
#
                         Petal.Length
#
                         Petal.Width
#
```

рса

Principal Component Analysis

Description

Principal Component Analysis

Usage

```
pca(data, vars, nFactorMethod = "parallel", nFactors = 1,
  minEigen = 1, rotation = "varimax", hideLoadings = 0.3,
  sortLoadings = FALSE, screePlot = FALSE, eigen = FALSE,
  factorCor = FALSE, factorSummary = FALSE, kmo = FALSE,
  bartlett = FALSE)
```

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Arguments

data the data as a data frame

vars a vector of strings naming the variables of interest in data

nFactorMethod 'parallel' (default), 'eigen' or 'fixed', the way to determine the number

of factors

nFactors an integer (default: 1), the number of components in the model

minEigen a number (default: 1), the minimal eigenvalue for a component to be included

in the model

rotation 'none', 'varimax' (default), 'quartimax', 'promax', 'oblimin', or 'simplimax',

the rotation to use in estimation

hideLoadings a number (default: 0.3), hide loadings below this value sortLoadings

TRUE or FALSE (default), sort the factor loadings by size

screePlot TRUE or FALSE (default), show scree plot

eigen TRUE or FALSE (default), show eigenvalue table factorCor TRUE or FALSE (default), show factor correlations factorSummary TRUE or FALSE (default), show factor summary

kmo TRUE or FALSE (default), show Kaiser-Meyer-Olkin (KMO) measure of sampling

adequacy (MSA) results

bartlett TRUE or FALSE (default), show Bartlett's test of sphericity results

Value

A results object containing:

results\$loadings a table a table results\$factorStats\$factorSummary results\$factorStats\$factorCor a table results\$modelFit\$fit a table results\$assump\$bartlett a table results\$assump\$kmo a table results\$eigen\$initEigen a table results\$eigen\$screePlot an image

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$loadings\$asDF as.data.frame(results\$loadings)

```
data('iris')
pca(iris, vars = vars(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width))
```

propTest2 43

```
#
  PRINCIPAL COMPONENT ANALYSIS
#
#
  Component Loadings
#
#
                1 Uniqueness
    Sepal.Length 0.890 0.2076
    Sepal.Width -0.460
#
                             0.7883
    Petal.Length 0.992
                             0.0168
#
    Petal.Width 0.965
                             0.0688
#
#
    Note. 'varimax' rotation was used
#
```

propTest2

Proportion Test (2 Outcomes)

Description

The Binomial test is used to test the Null hypothesis that the proportion of observations match some expected value. If the p-value is low, this suggests that the Null hypothesis is false, and that the true proportion must be some other value.

Usage

```
propTest2(data, vars, areCounts = FALSE, testValue = 0.5,
  hypothesis = "notequal", ci = FALSE, ciWidth = 95, bf = FALSE,
  priorA = 1, priorB = 1, ciBayes = FALSE, ciBayesWidth = 95,
  postPlots = FALSE)
```

data	the data as a data frame
vars	a vector of strings naming the variables of interest in data
areCounts	TRUE or FALSE (default), the variables are counts
testValue	a number (default: 0.5), the value for the null hypothesis
hypothesis	'notequal' (default), 'greater' or 'less', the alternative hypothesis
ci	TRUE or FALSE (default), provide confidence intervals
ciWidth	a number between 50 and 99.9 (default: 95), the confidence interval width
bf	TRUE or FALSE (default), provide Bayes factors
priorA	a number (default: 1), the beta prior 'a' parameter
priorB	a number (default: 1), the beta prior 'b' parameter
ciBayes	TRUE or FALSE (default), provide Bayesian credible intervals
ciBayesWidth	a number between 50 and 99.9 (default: 95), the credible interval width
postPlots	TRUE or FALSE (default), provide posterior plots

propTestN

Value

A results object containing:

```
results$table results$postPlots
```

a table of the proportions and test results an array of the posterior plots

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$table\$asDF as.data.frame(results\$table)

Examples

```
## Not run:
dat <- data.frame(x=c(8, 15))
propTest2(dat, vars = x, areCounts = TRUE)
  PROPORTION TEST (2 OUTCOMES)
#
#
  Binomial Test
#
#
       Level Count Total Proportion p
#
              8 23 0.348 0.210
#
   x 1
                       23
#
      2
                15
                                 0.652 0.210
#
    Note. Ha is proportion != 0.5
#
#
## End(Not run)
```

 ${\tt propTestN}$

Proportion Test (N Outcomes)

Description

The X^2 Goodness of fit test (not to be confused with the X^2 test of independence), tests the Null hypothesis that the proportions of observations match some expected proportions. If the p-value is low, this suggests that the Null hypothesis is false, and that the true proportions are different to those tested.

Usage

```
propTestN(data, var, counts = NULL, expected = FALSE, ratio = NULL,
    formula)
```

propTestN 45

Arguments

data	the data as a data frame
var	the variable of interest in data (not necessary when using a formula, see the examples)
counts	the counts in data
expected	TRUE or FALSE (default), whether expected counts should be displayed
ratio	a vector of numbers: the expected proportions
formula	(optional) the formula to use, see the examples

Value

A results object containing:

```
results$props a table of the proportions results$tests a table of the test results
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results props asDF

```
as.data.frame(results$props)
```

```
data('HairEyeColor')
dat <- as.data.frame(HairEyeColor)</pre>
propTestN(formula = Freq \sim Eye, data = dat, ratio = c(1,1,1,1))
  PROPORTION TEST (N OUTCOMES)
#
#
# Proportions
# -----
  Level Count Proportion
#
#
 -----
  Brown 220 0.372
Blue 215 0.363
Hazel 93 0.157
Green 64 0.108
#
#
# X<sup>2</sup> Goodness of Fit
#
  X² df p
# -----
  133 3 < .001
#
# -----
```

46 reliability

#

ility Analysis	Reliability	reliability
----------------	-------------	-------------

Description

Reliability Analysis

Usage

```
reliability(data, vars, alphaScale = TRUE, omegaScale = FALSE,
  meanScale = FALSE, sdScale = FALSE, corPlot = FALSE,
  alphaItems = FALSE, omegaItems = FALSE, meanItems = FALSE,
  sdItems = FALSE, itemRestCor = FALSE, revItems = NULL)
```

Arguments

data	the data as a data frame
vars	a vector of strings naming the variables of interest in data
alphaScale	TRUE (default) or FALSE, provide Cronbach's alpha
omegaScale	TRUE or FALSE (default), provide McDonald's omega
meanScale	TRUE or FALSE (default), provide the mean
sdScale	TRUE or FALSE (default), provide the standard deviation
corPlot	TRUE or FALSE (default), provide a correlation plot
alphaItems	TRUE or FALSE (default), provide what the Cronbach's alpha would be if the item was dropped
omegaItems	TRUE or FALSE (default), provide what the McDonald's omega would be if the item was dropped
meanItems	TRUE or FALSE (default), provide item means
sdItems	TRUE or FALSE (default), provide item standard deviations
itemRestCor	TRUE or FALSE (default), provide item-rest correlations
revItems	a vector containing strings naming the varibales that are reverse scaled

Value

A results object containing:

results\$scale	a table
results\$items	a table
results\$corPlot	an image

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```
Tables can be converted to data frames with asDF or as.data.frame. For example: results$scale$asDF as.data.frame(results$scale)
```

Examples

ttestIS

Independent Samples T-Test

Description

The Student's Independent samples t-test (sometimes called a two-samples t-test) is used to test the null hypothesis that two groups have the same mean. A low p-value suggests that the null hypothesis is not true, and therefore the group means are different.

Usage

```
ttestIS(data, vars, group, students = TRUE, bf = FALSE,
bfPrior = 0.707, welchs = FALSE, mann = FALSE,
hypothesis = "different", norm = FALSE, qq = FALSE, eqv = FALSE,
meanDiff = FALSE, effectSize = FALSE, ci = FALSE, ciWidth = 95,
desc = FALSE, plots = FALSE, miss = "perAnalysis", formula)
```

data	the data as a data frame
vars	the dependent variables (not necessary when using a formula, see the examples)
group	the grouping variable with two levels (not necessary when using a formula, see the examples)
students	TRUE (default) or FALSE, perform Student's t-tests

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bf TRUE or FALSE (default), provide Bayes factors

bfPrior a number between 0.5 and 2 (default 0.707), the prior width to use in calculating

Bayes factors

welchs TRUE or FALSE (default), perform Welch's t-tests

mann TRUE or FALSE (default), perform Mann-Whitney U tests

hypothesis 'different' (default), 'oneGreater' or 'twoGreater', the alternative hy-

pothesis; group 1 different to group 2, group 1 greater than group 2, and group

2 greater than group 1 respectively

norm TRUE or FALSE (default), perform Shapiro-Wilk tests of normality

qq TRUE or FALSE (default), provide Q-Q plots of residuals

eqv TRUE or FALSE (default), perform Levene's tests for equality of variances

meanDiff TRUE or FALSE (default), provide means and standard errors

effectSize TRUE or FALSE (default), provide effect sizes

ci TRUE or FALSE (default), provide confidence intervals

ciWidth a number between 50 and 99.9 (default: 95), the width of confidence intervals

desc TRUE or FALSE (default), provide descriptive statistics
plots TRUE or FALSE (default), provide descriptive plots

miss 'perAnalysis' or 'listwise', how to handle missing values; 'perAnalysis'

excludes missing values for individual dependent variables, 'listwise' ex-

cludes a row from all analyses if one of its entries is missing.

formula (optional) the formula to use, see the examples

Details

The Student's independent t-test assumes that the data from each group are from a normal distribution, and that the variances of these groups are equal. If unwilling to assume the groups have equal variances, the Welch's t-test can be used in it's place. If one is additionally unwilling to assume the data from each group are from a normal distribution, the non-parametric Mann-Whitney U test can be used instead (However, note that the Mann-Whitney U test has a slightly different null hypothesis; that the distributions of each group is equal).

Value

A results object containing:

results\$ttest a table containing the t-test results results\$assum\$norm a table containing the normality tests

results\$assum\$eqv a table containing the equality of variances tests

results\$desc a table containing the group descriptives

results\$plots an array of groups of plots

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$ttest\$asDF

ttestOneS 49

```
as.data.frame(results$ttest)
```

Examples

ttestOneS

One Sample T-Test

Description

The Student's One-sample t-test is used to test the null hypothesis that the true mean is equal to a particular value (typically zero). A low p-value suggests that the null hypothesis is not true, and therefore the true mean must be different from the test value.

Usage

```
ttestOneS(data, vars, students = TRUE, bf = FALSE, bfPrior = 0.707,
wilcoxon = FALSE, testValue = 0, hypothesis = "dt", norm = FALSE,
qq = FALSE, meanDiff = FALSE, effectSize = FALSE, ci = FALSE,
ciWidth = 95, desc = FALSE, plots = FALSE, miss = "perAnalysis",
mann = FALSE)
```

data	the data as a data frame
vars	a vector of strings naming the variables of interest in data
students	TRUE (default) or FALSE, perform Student's t-tests
bf	TRUE or FALSE (default), provide Bayes factors
bfPrior	a number between 0.5 and 2.0 (default 0.707), the prior width to use in calculating Bayes factors
wilcoxon	TRUE or FALSE (default), perform Wilcoxon signed rank tests
testValue	a number specifying the value of the null hypothesis

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'dt' (default), 'gt' or 'lt', the alternative hypothesis; different to testValue, hypothesis greater than testValue, and less than testValue respectively TRUE or FALSE (default), perform Shapiro-wilk tests of normality norm TRUE or FALSE (default), provide a Q-Q plot of residuals qq meanDiff TRUE or FALSE (default), provide means and standard deviations effectSize TRUE or FALSE (default), provide Cohen's d effect sizes TRUE or FALSE (default), provide confidence intervals for the mean difference ci ciWidth a number between 50 and 99.9 (default: 95), the width of confidence intervals TRUE or FALSE (default), provide descriptive statistics desc TRUE or FALSE (default), provide descriptive plots plots 'perAnalysis' or 'listwise', how to handle missing values; 'perAnalysis' miss

excludes missing values for individual dependent variables, 'listwise' ex-

cludes a row from all analyses if one of its entries is missing.

mann deprecated

Details

The Student's One-sample t-test assumes that the data are from a normal distribution – in the case that one is unwilling to assume this, the non-parametric Wilcoxon signed-rank can be used in it's place (However, note that the Wilcoxon signed-rank has a slightly different null hypothesis; that the *median* is equal to the test value).

Value

A results object containing:

results\$ttest a table containing the t-test results
results\$normality a table containing the normality test results
results\$descriptives a table containing the descriptives
results\$plots an image of the descriptive plots
results\$qq an array of Q-Q plots

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$ttest\$asDF as.data.frame(results\$ttest)

```
data('ToothGrowth')

ttestOneS(ToothGrowth, vars = vars(len, dose))

#
# ONE SAMPLE T-TEST
#
```

ttestPS 51

ttestPS

Paired Samples T-Test

Description

The Student's paired samples t-test (sometimes called a dependent-samples t-test) is used to test the null hypothesis that the difference between pairs of measurements is equal to zero. A low p-value suggests that the null hypothesis is not true, and that the difference between the measurement pairs is not zero.

Usage

```
ttestPS(data, pairs, students = TRUE, bf = FALSE, bfPrior = 0.707,
wilcoxon = FALSE, hypothesis = "different", norm = FALSE,
qq = FALSE, meanDiff = FALSE, effectSize = FALSE, ci = FALSE,
ciWidth = 95, desc = FALSE, plots = FALSE, miss = "perAnalysis")
```

data	the data as a data frame
pairs	a list of lists specifying the pairs of measurement in data
students	TRUE (default) or FALSE, perform Student's t-tests
bf	TRUE or FALSE (default), provide Bayes factors
bfPrior	a number between 0.5 and 2 (default 0.707), the prior width to use in calculating Bayes factors $\frac{1}{2}$
wilcoxon	TRUE or FALSE (default), perform Wilcoxon signed rank tests
hypothesis	'different' (default), 'oneGreater' or 'twoGreater', the alternative hypothesis; group 1 different to group 2, group 1 greater than group 2, and group 2 greater than group 1 respectively
norm	TRUE or FALSE (default), perform Shapiro-wilk normality tests
qq	TRUE or FALSE (default), provide a Q-Q plot of residuals
meanDiff	TRUE or FALSE (default), provide means and standard errors
effectSize	TRUE or FALSE (default), provide effect sizes
ci	TRUE or FALSE (default), provide confidence intervals

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ciWidth	a number between 50 and 99.9 (default: 95), the width of confidence intervals
desc	TRUE or FALSE (default), provide descriptive statistics
plots	TRUE or FALSE (default), provide descriptive plots
miss	'perAnalysis' or 'listwise', how to handle missing values; 'perAnalysis' excludes missing values for individual dependent variables, 'listwise' excludes a row from all analyses if one of its entries is missing

Details

The Student's paired samples t-test assumes that pair differences follow a normal distribution – in the case that one is unwilling to assume this, the non-parametric Wilcoxon signed-rank can be used in it's place (However, note that the Wilcoxon signed-rank has a slightly different null hypothesis; that the two groups of measurements follow the same distribution).

Value

A results object containing:

```
results$ttest a table containing the t-test results
results$norm a table containing the normality test results
results$desc a table containing the descriptives
results$plots an array of the descriptive plots
```

Tables can be converted to data frames with asDF or as.data.frame. For example: results\$ttest\$asDF as.data.frame(results\$ttest)

```
## Not run:
data('bugs', package = 'jmv')
ttestPS(bugs, pairs = list(
     list(i1 = 'LDLF', i2 = 'LDHF')))
  PAIRED SAMPLES T-TEST
#
  Paired Samples T-Test
#
#
                          statistic df p
#
   LDLF LDHF Student's t
                             -6.65 90.0 < .001
#
   _____
#
#
## End(Not run)
```

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