Package 'pubh'

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Title A Toolbox for Public Health and Epidemiology

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Description A toolbox for making R functions and capabilities more accessible to students and professionals from Epidemiology and Public Health related disciplines. Includes a function to report coefficients and confidence intervals from models using robust standard errors (when available), functions that expand 'ggplot2' plots and functions relevant for introductory papers in Epidemiology or Public Health. Please note that use of the provided data sets is for educational purposes only.

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```
axis_labs
```

Apply labels from variables to axis-labels in plots.

Description

axis_labs takes labels from labelled data to use them as axis-labels for plots generated by gformula or ggplot2.

Usage

axis_labs(object)

Arguments

object ggplot2 object (see examples).

Details

This functions is helpful when data has been already labelled by sjlabelled. It retrives variable labels and use them for plotting.

Value

A ggplot2 object.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm |>
var_labels(
    dl.milk = "Breast-milk intake (dl/day)",
    sex = "Sex",
    weight = "Child weight (kg)",
    ml.suppl = "Milk substitute (ml/day)",
    mat.weight = "Maternal weight (kg)",
    mat.height = "Maternal height (cm)"
)
kfm |>
    gf_point(weight ~ dl.milk) |>
    gf_lm(col = 2, interval = "confidence", col = 2) |>
```

```
axis_labs()
```

```
kfm |>
box_plot(dl.milk ~ sex, fill = "thistle", alpha = 0.8) |>
axis_labs() |>
gf_star(x1 = 1, y1 = 10.9, x2 = 2, y2 = 11, y3 = 11.2)
```

bar_error

Bar charts with error bars.

Description

bar_error constructs bar charts in with error bars showing 95 confidence intervals around mean values. High of bars represent mean values.

Usage

```
bar_error(
   object = NULL,
   formula = NULL,
   data = NULL,
   fill = "indianred3",
   col = "black",
   alpha = 0.7,
   ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $y \sim x$ or $y \sim x z$ where y is a numerical variable and both x and z are factors.
data	A data frame where the variables in the formula can be found.
fill	Colour used to fill the bars.
col	Colour used for the borders of the bars.
alpha	Opacity of the colour fill $(0 = invisible, 1 = opaque)$.
	Additional information passed to gf_summary.

Examples

```
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
data(birthwt, package = "MASS")
birthwt <- birthwt |>
   mutate(
```

Bernard

```
smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
Race = factor(race > 1, labels = c("White", "Non-white"))
) |>
var_labels(
bwt = "Birth weight (g)",
smoke = "Smoking status"
)
birthwt |>
bar_error(bwt ~ smoke, fill = "plum3")
birthwt |>
bar_error(bwt ~ smoke | Race, fill = "plum3")
birthwt |>
bar_error(bwt ~ smoke, fill = ~Race)
```

```
Bernard
```

Survival of patients with sepsis.

Description

A randomised, double-blind, placebo-controlled trial of intravenous ibuprofen in 455 patients who had sepsis, defined as fever, tachycardia, tachypnea, and acute failure of at least one organ system.

Usage

Bernard

Format

A labelled tibble with 455 rows and 9 variables:

id Patient ID

treat Treatment, factor with levels "Placebo" and "Ibuprofen".

race Race/ethnicity, factor with levels "White", "African American" and "Other".

fate Mortality status at 30 days, factor with levels "Alive" and "Dead".

apache Baseline APACHE score.

o2del Oxygen delivery at baseline.

followup Follow-up time in hours.

temp0 Baseline temperature in centigrades.

temp10 Temperature after 36 hr in centigrades.

Source

Bernard, GR, et al. (1997) The effects of ibuprofen on the physiology and survival of patients with sepsis, N Engl J Med 336: 912–918.

bland_altman

Description

Bland-Altman agreement plots.

Usage

```
bland_altman(
  object = NULL,
  formula = NULL,
  data = NULL,
  pch = 20,
  size = 1,
  col = "black",
  transform = FALSE,
  ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $y \sim x$ (see details).
data	A data frame where the variables in the formula can be found.
pch	Symbol for plotting data.
size	Size of the symbol using to plot data.
col	Colour used for the symbol to plot data.
transform	Logical, should ratios instead of difference be used to construct the plot?
	Further arguments passed to gf_point.

Details

bland_altman constructs Bland-Altman agreement plots.

Variables in formula are continuous paired observations. When the distribution of the outcome is not normal, but becomes normal with a log-transformation, bland_altman can plot the ratio between outcomes (difference in the log scale) by using option transform = TRUE.

Examples

data(wright, package = "ISwR")

wright |>
 bland_altman(mini.wright ~ std.wright,

box_plot

```
pch = 16,
ylab = "Large-mini expiratory flow rate (l/min)",
xlab = "Mean expiratory flow rate (l/min)"
) |>
gf_labs(
y = "Large-mini expiratory flow rate (l/min)",
x = "Mean expiratory flow rate (l/min)"
)
data(Sharples)
Sharples |>
bland_altman(srweight ~ weight, transform = TRUE) |>
gf_labs(x = "Mean of weights (kg)", y = "Measured weight / Self-reported weight")
```

```
box_plot
```

Construct box plots.

Description

box_plot is a wrap function that calls gf_boxplot to construct more aesthetic box plots.

Usage

```
box_plot(
  object = NULL,
  formula = NULL,
  data = NULL,
  fill = "indianred3",
  alpha = 0.7,
  outlier.shape = 20,
  outlier.size = 1,
  ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $y \sim x$ where y is a numerical variable and x is a factor.
data	A data frame where the variables in the formula can be found.
fill	Colour used for the box passed to gf_boxplot.
alpha	Opacity ($0 = invisible$, $1 = opaque$).
outlier.shape	Shape (pch) used as symbol for the outliers.
outlier.size	Size of the outlier symbol.
	Further arguments passed to gf_boxplot.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm l>
  var_labels(
   dl.milk = "Breast-milk intake (dl/day)",
   sex = "Sex",
   weight = "Child weight (kg)",
   ml.suppl = "Milk substitute (ml/day)",
   mat.weight = "Maternal weight (kg)",
   mat.height = "Maternal height (cm)"
  )
kfm |>
  box_plot(dl.milk ~ sex, fill = "thistle", alpha = 0.8)
t.test(dl.milk ~ sex, data = kfm)
kfm |>
  box_plot(dl.milk ~ sex, fill = "thistle", alpha = 0.8) |>
  gf_star(1, 10.9, 2, 11, 11.4, legend = "p = 0.035", size = 2.5)
```

Brenner

Prevalence of Helicobacter pylori infection in preschool children.

Description

A data set containing the prevalence of Helicobacter pylori infection in preschool children according to parental history of duodenal or gastric ulcer.

Usage

Brenner

Format

A labelled tibble with 863 rows and 2 variables:

ulcer History of duodenal or gastric ulcer, factor with levels "No" and "Yes".

infected Infected with Helicobacter pylori, factor with levels "No" and "Yes".

Source

Brenner H, Rothenbacher D, Bode G, Adler G (1998) Parental history of gastric or duodenal ulcer and prevalence of Helicobacter pylori infection in preschool children: population based study. BMJ 316:665.

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Description

bst estimates confidence intervals around the mean, median or geo_mean.

Usage

bst(x, stat = "mean", n = 1000, CI = 95, digits = 2)

Arguments

х	A numerical variable. Missing observations are removed by default.
stat	Statistic, either "mean" (default), "median" or "gmean" (geometric mean).
n	Number of replicates for the bootstrap (n=1000 by default).
CI	Confidence intervals (CI=95 by default).
digits	Number of digits for rounding (default = 2).

Value

A data frame with the estimate and confidence intervals.

Examples

```
data(IgM, package = "ISwR")
bst(IgM, "median")
bst(IgM, "gmean")
```

coef_det

Coefficient of determination.

Description

coef_det estimates the coefficient of determination (r-squared) from fitted (predicted) and observed values. Outcome from the model is assumed to be numerical.

Usage

coef_det(obs, fit)

Arguments

obs	Vector with observed values (numerical outcome).
fit	Vector with fitted (predicted) values.

bst

estat

Value

A scalar, the coefficient of determination (r-squared).

Examples

```
## Linear regression:
Riboflavin <- seq(0, 80, 10)
OD <- 0.0125 * Riboflavin + rnorm(9, 0.6, 0.03)
titration <- data.frame(Riboflavin, OD)</pre>
model1 <- lm(OD ~ Riboflavin, data = titration)</pre>
summary(model1)
coef_det(titration$OD, fitted(model1))
## Non-linear regression:
library(nlme, quietly = TRUE)
data(Puromycin)
mm.tx <- gnls(rate ~ SSmicmen(conc, Vm, K),</pre>
  data = Puromycin,
  subset = state == "treated"
)
summary(mm.tx)
coef_det(Puromycin$rate[1:12], mm.tx$fitted)
```

estat

Descriptive statistics for continuous variables.

Description

estat calculates descriptives of numerical variables.

Usage

```
estat(object = NULL, formula = NULL, data = NULL, digits = 2, label = NULL)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $\sim x$ or $\sim x z$ (for groups).
data	A data frame where the variables in the formula can be found.
digits	Number of digits for rounding (default = 2).
label	Label used to display the name of the variable (see examples).

Value

A data frame with descriptive statistics.

expand_df

See Also

summary.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm |>
  var_labels(
   dl.milk = "Breast-milk intake (dl/day)",
   sex = "Sex",
   weight = "Child weight (kg)",
   ml.suppl = "Milk substitute (ml/day)",
   mat.weight = "Maternal weight (kg)",
   mat.height = "Maternal height (cm)"
  )
kfm |>
  estat(~dl.milk)
estat(~ dl.milk | sex, data = kfm)
kfm |>
  estat(~ weight | sex)
```

expand_df <i>Expand a data frame</i> .
--

Description

expand_df expands a data frame by a vector of frequencies.

Usage

```
expand_df(aggregate.data, index.var = "Freq", retain.freq = FALSE)
```

Arguments

aggregate.data	A data frame.
index.var	A numerical variable with the frequencies (counts).
retain.freq	Logical expression indicating if frequencies should be kept.

Details

This is a generic function that resembles weighted frequencies in other statistical packages (for example, Stata). expand.df was adapted from a function developed by deprecated package epicalc (now package epiDisplay).

Value

An expanded data frame with replicates given by the frequencies.

Examples

```
Freq <- c(5032, 5095, 41, 204)
Mortality <- gl(2, 2, labels = c("No", "Yes"))
Calcium <- gl(2, 1, 4, labels = c("No", "Yes"))
anyca <- data.frame(Freq, Mortality, Calcium)
anyca
anyca.exp <- expand_df(anyca)
with(anyca.exp, table(Calcium, Mortality))</pre>
```

Fentress

Migraine pain reduction.

Description

Randomised control trial on children suffering from frequent and severe migraine. Control group represents untreated children. The active treatments were either relaxation alone or relaxation with biofeedback.

Usage

Fentress

Format

A labelled tibble with 18 rows and 2 variables:

pain Reduction in weekly headache activity expressed as percentage of baseline data.

group Group, a factor with levels "Untreated", "Relaxation" (alone) and "Biofeedback" (relaxation and biofeedback).

Source

Fentress, DW, et al. (1986) Biofeedback and relaxation-response in the treatment of pediatric migraine. Dev Med Child Neurol 28:1 39-46.

Altman, DA (1991) Practical statistics for medical research. Chapman & Hall/CRC.

Examples

```
data(Fentress)
```

```
Fentress |>
   strip_error(pain ~ group)
```

freq_cont

Description

freq_cont tabulates a continuous variable by given classes.

Usage

 $freq_cont(x, bks, dg = 2)$

Arguments

x	A numerical (continuous) variable. Ideally, relatively long (greater than 100 observations).
bks	Breaks defining the classes (see example).
dg	Number of digits for rounding (default = 2).

Value

A data frame with the classes, the mid-point, the frequencies, the relative and cumulative frequencies.

Examples

```
data(IgM, package = "ISwR")
Ab <- data.frame(IgM)
estat(~IgM, data = Ab)
freq_cont(IgM, seq(0, 4.5, 0.5))</pre>
```

gen_bst_df

Generate a data frame with estimate and bootstrap CIs.

Description

gen_bst_df is a function called that generates a data frame with confidence intervals of a continuous variable by levels of one or two categorical ones (factors).

Usage

```
gen_bst_df(object = NULL, formula = NULL, data = NULL, stat = "mean", ...)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $y \sim x$ or $y \sim x z$ where y is a numerical variable and both x and z are factors.
data	A data frame where the variables in the formula can be found.
stat	Statistic used for bst.
	Passes optional arguments to bst.

Value

A data frame with the confidence intervals by level.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm |>
 var_labels(
   dl.milk = "Breast-milk intake (dl/day)",
   sex = "Sex",
   weight = "Child weight (kg)",
   ml.suppl = "Milk substitute (ml/day)",
   mat.weight = "Maternal weight (kg)",
   mat.height = "Maternal height (cm)"
  )
kfm |>
  gen_bst_df(dl.milk ~ sex)
data(birthwt, package = "MASS")
require(dplyr, quietly = TRUE)
birthwt <- mutate(birthwt,</pre>
  smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
  Race = factor(race > 1, labels = c("White", "Non-white"))
)
birthwt <- birthwt |>
  var_labels(
   bwt = "Birth weight (g)",
   smoke = "Smoking status"
  )
gen_bst_df(bwt ~ smoke | Race, data = birthwt)
```

geo_mean

Geometric mean.

Description

Geometric mean.

Usage

geo_mean(x)

Arguments

х

A numeric variable with no negative values.

Value

A scalar, the calculated geometric mean.

Examples

```
data(IgM, package = "ISwR")
Ab <- data.frame(IgM)
estat(~IgM, data = Ab)
geo_mean(IgM)</pre>
```

get_r2

Estimate R2 or Pseudo-R2 from regression models

Description

get_r2 is a is a wrap function that calls r2 from package performance. Calculates the R2 or pseudo-R2 value for different regression model objects, returning a character object for easy printing in tables of coefficients.

Usage

get_r2(model, ...)

Arguments

model	A statistical regression model.
	Additional arguments passed to r2.

Details

The main purpose of get_r2 is to allow easy printing of R2 value in tables of coefficients (see examples).

See Also

r2.

Examples

```
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)

data(birthwt, package = "MASS")
birthwt <- birthwt |>
  mutate(
    smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
    race = factor(race, labels = c("White", "African American", "Other"))
    ) |>
    var_labels(
    bwt = "Birth weight (g)",
    smoke = "Smoking status",
    race = "Race"
    )

model_norm <- lm(bwt ~ smoke + race, data = birthwt)
get_r2(model_norm)</pre>
```

gf_star

Annotating a plot to display differences between groups.

Description

gf_star Is a function used to display differences between groups (see details).

Usage

gf_star(fig, x1, y1, x2, y2, y3, legend = "*", ...)

Arguments

fig	A gformula object.
x1	Position in x for the start of the horizontal line.
y1	Position in y for the start of the vertical line, below to the horizontal line.
x2	Position in x for the end of the horizontal line.
y2	Position in y where the horizontal line is drawn.

у3	Position in y where the text is added.
legend	Character text used for annotating the plot.
	Additional information passed to gf_text.

Details

This function draws an horizontal line from coordinate (x1, y2) to coordinate (x2, y2). Draws vertical lines below the horizontal line, towards data, from (x1, y1) to (x1, y2) and from (x2, y1) to (x2, y2). Finally, adds text above the horizontal line, at the mid point between x1 and x2. See examples.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm |>
 var_labels(
   dl.milk = "Breast-milk intake (dl/day)",
   sex = "Sex",
   weight = "Child weight (kg)",
   ml.suppl = "Milk substitute (ml/day)",
   mat.weight = "Maternal weight (kg)",
   mat.height = "Maternal height (cm)"
 )
kfm |>
 box_plot(dl.milk ~ sex, fill = "thistle", alpha = 0.8) |>
 gf_star(x1 = 1, y1 = 10.9, x2 = 2, y2 = 11, y3 = 11.2)
kfm |>
 box_plot(dl.milk ~ sex, fill = "thistle", alpha = 0.8) |>
 gf_star(1, 10.9, 2, 11, 11.4, legend = "p = 0.035", size = 2.5)
data(energy, package = "ISwR")
energy <- energy |>
 var_labels(
   expend = "Energy expenditure (MJ/day)",
   stature = "Stature"
 )
energy |>
 strip_error(expend ~ stature, col = "red") |>
 gf_star(1, 13, 2, 13.2, 13.4, "**")
```

Description

glm_coef displays estimates with confidence intervals and p-values from generalised linear models (see Details).

Usage

```
glm_coef(
  model,
  digits = 2,
  alpha = 0.05,
  labels = NULL,
  se_rob = FALSE,
  type = "cond",
  exp_norm = FALSE
)
```

Arguments

model	A model from any of the classes listed in the details section.
digits	A scalar, number of digits for rounding the results (default = 2).
alpha	Significant level (default = 0.05) used to calculate confidence intervals.
labels	An optional character vector with the names of the coefficients (including inter- cept).
se_rob	Logical, should robust errors be used to calculate confidence intervals? (default = FALSE).
type	Character, either "cond" (condensed) or "ext" (extended). See details.
exp_norm	Logical, should estimates and confidence intervals should be exponentiated? (for family == "gaussian").

Details

glm_coef recognises objects (models) from the following classes: clm, clogit, coxph, gee, glm, glmerMod, lm, lme, lmerMod, multinom, negbin, polr and surveg

For models from logistic regression (including conditional logistic, ordinal and multinomial), Poisson or survival analysis, coefficient estimates and corresponding confidence intervals are automatically exponentiated (back-transformed).

By default, glm_coef uses naive standard errors for calculating confidence intervals but has the option of using robust standard errors instead.

glm_coef can display two different data frames depending on the option of type, for type type = "cond" (the default), glm_coef displays the standard table of coefficients with confidence intervals and p-values; for type = "ext", glm_coef displays additional statistics including standard errors.

Please read the Vignette on Regression for more details.

Value

A data frame with estimates, confidence intervals and p-values from glm objects.

harm_mean

Examples

```
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
## Continuous outcome.
data(birthwt, package = "MASS")
birthwt <- birthwt |>
  mutate(
   smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
   race = factor(race, labels = c("White", "African American", "Other"))
  ) |>
  var_labels(
   bwt = "Birth weight (g)",
   smoke = "Smoking status",
   race = "Race"
  )
model_norm <- lm(bwt ~ smoke + race, data = birthwt)</pre>
glm_coef(model_norm, labels = model_labels(model_norm))
## Logistic regression.
data(diet, package = "Epi")
model_binom <- glm(chd ~ fibre, data = diet, family = binomial)</pre>
model_binom |>
  glm_coef(labels = c("Constant", "Fibre intake (g/day)"))
model_binom |>
  glm_coef(labels = c("Constant", "Fibre intake (g/day)"), type = "ext")
```

harm_mean

Description

Harmonic mean.

Usage

```
harm_mean(x)
```

Arguments ×

A numeric variable with no zero values.

Harmonic mean.

Value

A scalar, the calculated harmonic mean.

Examples

```
data(IgM, package = "ISwR")
Ab <- data.frame(IgM)
estat(~IgM, data = Ab)
harm_mean(IgM)</pre>
```

hist_norm

Histogram with Normal density curve.

Description

hist_norm constructs histograms and adds corresponding Normal density curve.

Usage

```
hist_norm(
   object = NULL,
   formula = NULL,
   data = NULL,
   bins = 20,
   fill = "indianred3",
   color = "black",
   alpha = 0.4,
   ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $\ y $ or $\ y x $ where y is a numerical variable and x is a factor.
data	A data frame where the variables in the formula can be found.
bins	Number of bins of the histogram.
fill	Colour to fill the bars of the histogram.
color	Colour used for the border of the bars.
alpha	Opacity ($0 = invisible$, $1 = opaque$).
	Further arguments passed to gf_dhistogram.

Hodgkin

Examples

```
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
data(birthwt, package = "MASS")
birthwt <- birthwt |>
  mutate(
    smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
   Race = factor(race > 1, labels = c("White", "Non-white"))
  ) |>
  var_labels(
   bwt = "Birth weight (g)",
   smoke = "Smoking status"
  )
birthwt |>
  hist_norm(~bwt, alpha = 0.7, bins = 20, fill = "cadetblue")
birthwt |>
  hist_norm(~ bwt | smoke, alpha = 0.7, bins = 20, fill = "cadetblue")
```

Hodgkin

T-cell counts from Hodgkin's disease patients.

Description

Number of CD4+ T-cells and CD8+ T-cells in blood samples from patients in remission from Hodgkin's disease or in remission from disseminated malignancies.

Usage

Hodgkin

Format

A labelled tibble with 40 rows and 3 variables:

CD4 Concentration of CD4+ T-cells (cells / mm³).

CD8 Concentration of CD8+ T-cells (cells / mm^3).

Group Group, factor with levels "Non-Hodgkin" and "Hodgkin".

Source

Shapiro, CM, et al (1986) Immunologic status of patients in remission from Hodgkin's disease and disseminated malignancies. Am J Med Sci 293:366-370.

Altman, DA (1991) Practical statistics for medical research. Chapman & Hall/CRC.

Examples

```
data(Hodgkin)
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
Hodgkin <- Hodgkin |>
  mutate(
   Ratio = CD4 / CD8
  ) |>
  var_labels(
   Ratio = "CD4+ / CD8+ T-cells"
  )
estat(~ Ratio | Group, data = Hodgkin)
Hodgkin |>
  qq_plot(~ Ratio | Group)
Hodgkin$Ratio <- Hodgkin$CD4 / Hodgkin$CD8</pre>
estat(~ Ratio | Group, data = Hodgkin)
qq_plot(~ Ratio | Group, data = Hodgkin)
```

|--|--|

Description

inv_logit Calculates the inverse of the logit (probability in logistic regression)

Usage

inv_logit(x)

Arguments

Х

Numerical value used to compute the inverse of the logit.

jac			

Ranks leverage observations from Jackknife method.

Description

jack_knife Ranks the squared differences between mean values from Jackknife analysis (arithmetic mean estimated by removing one observation at a time) and the original mean value.

Kirkwood

Usage

jack_knife(x)

Arguments

х

A numeric variable. Missing values are removed by default.

Value

Data frame with the ranked squared differences.

See Also

rank_leverage.

Examples

```
x <- rnorm(10, 170, 8)
x
mean(x)
jack_knife(x)
x <- rnorm(100, 170, 8)
mean(x)
head(jack_knife(x))</pre>
```

Kirkwood

Body weight and plasma volume.

Description

Body weight and plasma volume in eight healthy men.

Usage

Kirkwood

Format

A labelled data frame with 8 rows and 3 variables:

subject Subject ID.

weight Body weight in kg.

volume Plasma volume in litres.

Source

Kirkwood, BR and Sterne, JAC (2003) Essential Medical Statistics. Second Edition. Blackwell.

Examples

data(Kirkwood)

```
Kirkwood |>
 gf_point(volume ~ weight) |>
 gf_lm(col = "indianred3", interval = "confidence", fill = "indianred3")
```

knife_mean Jackknife for means.

Description

knife_mean is an internal function. Calculates arithmetic means by removing one observation at a time.

Usage

knife_mean(x)

Arguments

Х

A numerical variable. Missing values are removed for the mean calculation.

Value

A vector with the mean calculations.

Examples

```
x <- rnorm(10, 170, 8)
x
mean(x)
knife_mean(x)</pre>
```

leverage	Leverage.		
----------	-----------	--	--

Description

leverage is an internal function called by rank_leverage.

Usage

leverage(x)

logistic_gof

Arguments

х

A numeric variable. Missing values are removed by default.

Details

Estimates the leverage of each observation around the arithmetic mean.

Value

Variable with corresponding leverage estimations

Examples

```
x <- rnorm(10, 170, 8)
x
mean(x)
leverage(x)
rank_leverage(x)</pre>
```

logistic_gof Goodness of fit for Logistic Regression.

Description

logistic_gof performs the Hosmer and Lemeshow test to test the goodness of fit of a logistic regression model. This function is part of residuals.lrm from package rms.

Usage

logistic_gof(model)

Arguments

model A logistic regression model object.

Author(s)

Frank Harell, Vanderbilt University <f.harrell@vanderbilt.edu>

References

Hosmer DW, Hosmer T, Lemeshow S, le Cessie S, Lemeshow S. A comparison of goodness-of-fit tests for the logistic regression model. Stat in Med 16:965–980, 1997.

Examples

```
data(diet, package = "Epi")
model <- glm(chd ~ fibre, data = diet, family = binomial)
glm_coef(model, labels = c("Constant", "Fibre intake (g/day)"))
logistic_gof(model)</pre>
```

Macmahon

Description

An international case-control study to test the hypothesis that breast cancer is related to the age that a woman gives childbirth.

Usage

Macmahon

Format

A labelled tibble with 185 rows and 2 variables:

cancer Diagnosed with breast cancer, a factor with levels "No" and "Yes".

age Age mother gives childbirth, factor with levels "<20", "20-24", "25-29", "30-34" and ">34".

Source

Macmahon, B. et al. (1970). Age at first birth and breast cancer risk. Bull WHO 43, 209-221.

mhor

Mantel-Haenszel odds ratio.

Description

mhor computes odds ratios by levels of the stratum variable as well as the Mantel-Haenszel pooled odds ratio. The test for effect modification (test for interaction) is also displayed.

Usage

```
mhor(object = NULL, formula = NULL, data = NULL)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain.
	Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: outcome ~ stratum/exposure.
data	A data frame containing the variables used in formula.

Value

Odds ratios with 95 outcome by levels of stratum. The Mantel-Haenszel pooled OR and the test for effect modification is also reported.

model_labels

Examples

```
data(oswego, package = "epitools")
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
oswego <- oswego |>
  mutate(
   ill = factor(ill, labels = c("No", "Yes")),
   sex = factor(sex, labels = c("Female", "Male")),
   chocolate.ice.cream = factor(chocolate.ice.cream, labels = c("No", "Yes"))
  ) |>
  var_labels(
   ill = "Developed illness",
   sex = "Sex",
   chocolate.ice.cream = "Consumed chocolate ice cream"
  )
oswego |>
  mhor(ill ~ sex / chocolate.ice.cream)
```

```
model_labels
```

Using labels as coefficient names in tables of coefficients.

Description

model_labels replaces row names in glm_coef with labels from the original data frame.

Usage

```
model_labels(model, intercept = TRUE)
```

Arguments

model	A generalised linear model.
intercept	Logical, should the intercept be added to the list of coefficients?

Details

model_labels does not handle yet interaction terms, see examples. Please read the Vignette on Regression for more examples.

Examples

```
require(dplyr, quietly = TRUE)
require(sjlabelled, quietly = TRUE)
data(birthwt, package = "MASS")
birthwt <- birthwt |>
   mutate(
```

```
smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
   race = factor(race, labels = c("White", "African American", "Other"))
  ) |>
  var_labels(
   bwt = "Birth weight (g)",
   smoke = "Smoking status",
   race = "Race"
  )
model_norm <- lm(bwt ~ smoke + race, data = birthwt)</pre>
glm_coef(model_norm, labels = model_labels(model_norm))
model_int <- lm(formula = bwt ~ smoke * race, data = birthwt)</pre>
model_int |>
  glm_coef(labels = c(
   model_labels(model_int),
    "Smoker: African American",
    "Smoker: Other"
  ))
```

multiple

Multiple comparisons with plot.

Description

multiple displays results from post-doc analysis and constructs corresponding plot.

Usage

```
multiple(
  model,
  formula,
  adjust = "mvt",
  type = "response",
  reverse = TRUE,
  level = 0.95,
  digits = 2,
  ...
)
```

Arguments

model	A fitted model supported by emmeans, such as the result of a call to aov, lm , glm , etc.
formula	A formula with shape: $\sim y$ or $\sim y x$ (for interactions). Where y is the term of the model on which comparisons are made and x is a term interacting with y.

odds_trend

adjust	Method to adjust CIs and p-values (see details).
type	Type of prediction (matching "linear.predictor", "link", or "response").
reverse	Logical argument. Determines the direction of comparisons.
level	Confidence interval significance level.
digits	Number of digits for rounding (default = 2).
	Further arguments passed to emmeans.

Details

The default adjusting method is "mvt" which uses the multivariate t distribution. Other options are: "bonferroni", "holm", "hochberg", "tukey" and "none". The default option for argument reverse is to make reverse comparisons, i.e., against the reference level matching comparisons from 1m and g1m.

Value

A list with objects: df A data frame with adjusted p-values, fig_ci a plot with estimates and adjusted confidence intervals and fig_pval a plot comparing adjusted p-values.

See Also

emmeans, pwpp.

Examples

```
data(birthwt, package = "MASS")
birthwt$race <- factor(birthwt$race, labels = c("White", "African American", "Other"))
model_1 <- aov(bwt ~ race, data = birthwt)
multiple(model_1, ~race)$df
multiple(model_1, ~race)$fig_ci |>
gf_labs(y = "Race", x = "Difference in birth weights (g)")
multiple(model_1, ~race)$fig_pval |>
gf_labs(y = "Race")
```

odds_trend

Function to calculate OR using Wald CI, and plot trend.

Description

odds_trend calculates the odds ratio with confidence intervals (Wald) for different levels (three or more) of the exposure variable, constructs the corresponding plot and calculates if the trend is significant or not.

Usage

odds_trend(formula, data, angle = 45, hjust = 1, method = "wald", ...)

Arguments

formula	A formula with shape: outcome ~ exposure.
data	A data frame where the variables in the formula can be found.
angle	Angle of for the x labels (default = 45).
hjust	Horizontal adjustment for x labels (default = 1).
method	Method for calculating confidence interval around odds ratio.
	Passes optional arguments to oddsratio.

Details

odds_trend is a wrap function that calls oddsratio from package epitools. Additional methods for confidence intervals include: "midp", "fisher", and "small".

Value

A list with components df a data frame with the results and fig corresponding plot.

See Also

oddsratio.

Examples

A cross-sectional study looked at the association between obesity and a biopsy resulting
from mammography screening.

```
Freq <- c(3441, 34, 39137, 519, 20509, 280, 12149, 196, 11882, 199)
Biopsy <- gl(2, 1, 10, labels = c("No", "Yes"))
Weight <- gl(5, 2, 10, labels = c(
    "Underweight", "Normal", "Over (11-24%)",
    "Over (25-39%)", "Over (> 39%)"
))
breast <- data.frame(Freq, Biopsy, Weight)
breast
breast <- expand_df(breast)
require(sjlabelled, quietly = TRUE)
breast <- var_labels(breast,
    Weight = "Weight group"
))
odds_trend(Biopsy ~ Weight, data = breast)$df
odds_trend(Biopsy ~ Weight, data = breast)$fig</pre>
```

Description

Study of onchocerciasis ("river blindness") in Sierra Leone, in which subjects were classified according to whether they lived in villages in savannah or rainforest area.

Usage

Oncho

Format

A labelled tibble with 1302 rows and 7 variables:

id Subject ID.

mf Infected with Onchocerciasis volvulus, factor with levels "Not-infected" and "Infected".

area Area of residence, factor with levels "Savannah" and "Rainforest".

agegrp Age group in years, factor with levels "5-9", "10-19", "20-39" and "40+".

sex Subject sex, factor with levels "Male" and "Female".

mfload Microfiliariae load.

lesions Severe eye lesions, factor with levels "No" and "Yes".

Source

McMahon, JE, Sowa, SIC, Maude, GH and Kirkwood BR (1988) Onchocerciasis in Sierra Leone 2: a comparison of forest and savannah villages. Trans Roy Soc Trop Med Hyg 82: 595-600.

Kirkwood, BR and Sterne, JAC (2003) Essential Medical Statistics. Second Edition. Blackwell.

predict_inv Given y solve for x in a simple linear model.

Description

predict_inv Calculates the value the predictor x that generates value y with a simple linear model.

Usage

predict_inv(model, y)

prop_or

Arguments

model	A simple linear model object (class lm).
У	A numerical scalar, the value of the outcome for which we want to calculate the predictor x.

Value

The estimated value of the predictor.

Examples

Spectrophotometry example. Titration curve for riboflavin (nmol/ml). The sample has an absorbance ## of 1.15. Aim is to estimate the concentration of riboflavin in the sample.

```
Riboflavin <- seq(0, 80, 10)
OD <- 0.0125 * Riboflavin + rnorm(9, 0.6, 0.03)
titration <- data.frame(Riboflavin, OD)</pre>
require(sjlabelled, quietly = TRUE)
titration <- titration |>
 var_labels(
   Riboflavin = "Riboflavin (nmol/ml)",
    OD = "Optical density"
 )
titration |>
 gf_point(OD ~ Riboflavin) |>
 gf_smooth(col = "indianred3", se = TRUE, lwd = 0.5, method = "loess")
## Model with intercept different from zero:
model <- lm(OD ~ Riboflavin, data = titration)</pre>
glm_coef(model)
predict_inv(model, 1.15)
```

prop_or

Proportion, p1 from proportion p2 and OR.

Description

prop_or is a simple function to calculate a proportion, from another proportion and the odds ratio between them.

Usage

prop_or(p2, or)

pseudo_r2

Arguments

p2	The value of a proportion in the unexposed group (p2).
or	The odds ratio of p1/p2.

Value

p1, the proportion in the exposed group (p1).

Examples

```
flu <- matrix(c(20, 80, 220, 140), nrow = 2)
colnames(flu) <- c("Yes", "No")
rownames(flu) <- c("Vaccine", "Placebo")
flu
or <- (20 * 140) / (80 * 220)
p2 <- 80 / 220
prop_or(p2 = p2, or = or)
20 / 240</pre>
```

pseudo_r2	Pseudo R2 (logistic regression) pseudo_r2 Calculates R2 analogues
	(pseudo R2) of logistic regression.

Description

Pseudo R2 (logistic regression) pseudo_r2 Calculates R2 analogues (pseudo R2) of logistic regression.

Usage

pseudo_r2(model)

Arguments

model A logistic regression model.

Details

pseudo_r2 calculates three pseudo R2 of logistic regression models: 1) Nagelkerke, @0 Cox and Snell, 3) Hosmer and Lemeshow.

Value

A data frame with the calculated pseudo R2 values.

Examples

```
data(Oncho)
model_oncho <- glm(mf ~ area, data = Oncho, binomial)
glm_coef(model_oncho, labels = c("Constant", "Area (rainforest/savannah)"))
pseudo_r2(model_oncho)</pre>
```

```
qq_plot
```

Quantile-quantile plots against the standard Normal distribution.

Description

qq_plot constructs quantile-quantile plots against the standard normal distribution (also known as quantile-normal plots).

Usage

```
qq_plot(
  object = NULL,
  formula = NULL,
  data = NULL,
  pch = 20,
  col = "indianred3",
  ylab = NULL,
  ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $\sim x$ or $\sim x z$ where x is a numerical variable and z is a factor.
data	A data frame where the variables in the formula can be found.
pch	Point character passed to gf_qq.
col	Colour of the reference line, passed to gf_line.
ylab	Optional character passed as label for the y-axis.
	Further arguments passed to gf_qq.

Examples

```
data(kfm, package = "ISwR")
require(sjlabelled, quietly = TRUE)
kfm <- kfm |>
var_labels(
    dl.milk = "Breast-milk intake (dl/day)",
    sex = "Sex",
```

rank_influence

```
weight = "Child weight (kg)",
ml.suppl = "Milk substitute (ml/day)",
mat.weight = "Maternal weight (kg)",
mat.height = "Maternal height (cm)"
)
kfm |>
qq_plot(~dl.milk)
qq_plot(~ dl.milk | sex, data = kfm)
```

rank_influence Ranks observations based upon influence measures on models.

Description

rank_influence calculates influence measures of each data observation on models and then ranks them.

Usage

```
rank_influence(model)
```

Arguments

model A generalised linear model object.

Details

rank_influence is a wrap function that calls influence.measures, ranks observations on their significance influence on the model and displays the 10 most influential observations (if they are significant).

See Also

influence.measures.

Examples

```
data(diet, package = "Epi")
model <- glm(chd ~ fibre, data = diet, family = binomial)
rank_influence(model)</pre>
```

rank_leverage

Description

rank_leverage ranks observations by their leverage (influence) on the arithmetic mean.

Usage

```
rank_leverage(x)
```

Arguments

х

A numeric variable. Missing values are removed by default.

Value

A data frame ranking observations by their leverage around the mean.

See Also

jack_knife.

Examples

```
x <- rnorm(10, 170, 8)
x
mean(x)
rank_leverage(x)
x <- rnorm(100, 170, 8)
mean(x)
head(rank_leverage(x))</pre>
```

reference_range Reference range (reference interval).

Description

reference_range estimates the reference range (reference interval) of a numerical variable.

Usage

```
reference_range(avg, std)
```
rel_dis

Arguments

avg	The arithmetic mean (a scalar numerical value).
std	The standard deviation (a scalar numerical value).

Details

The reference range assumes normality and represents the limits that would include 95 observations.

Value

A data frame with the reference range limits.

Examples

```
x <- rnorm(100, 170, 8)
round(mean(x), 2)
round(sd(x), 2)
round(reference_range(mean(x), sd(x)), 2)</pre>
```

rel_dis

Relative Dispersion.

Description

Calculates the coefficient of variation (relative dispersion) of a variable. The relative dispersion is defined as the standard deviation over the arithmetic mean.

Usage

rel_dis(x)

Arguments

х

A numerical variable. NA's observations are removed by default.

Value

The coefficient of variation (relative dispersion).

Examples

height <- rnorm(100, 170, 8)
rel_dis(height)</pre>

Roberts

Description

A clinical trial on the value of extracorporeal membrane oxygenation for term neonates with severe respiratory failure. RCT compares active treatment against conventional management.

Usage

Roberts

Format

A labelled tibble with 185 rows and 2 variables:

emo Extracorporeal membrane oxygenation treatment, factor with levels "No" and "Yes". **survived** One year survival, factor with levels "No" and "Yes".

Source

Roberts, TE (1998) Extracorporeal Membrane Oxygenation Economics Working Group. Economic evaluation and randomised controlled trial of extracorporeal membrane oxygenation: UK collaborative trial. Brit Med J 317:911-16.

Rothman

Oral contraceptives and stroke.

Description

A case-control study of oral contraceptives and stroke in young women with presence or absence of hypertension. Cases represent thrombotic stroke and controls are hospital controls. The group of no hypertension includes normal blood pressure (<140/90 mm Hg) and borderline hypertension (140-159/90-94 mm Hg). Hypertension group includes moderate hypertension (160-179/95-109 mm Hg) and severe hypertension (180+/110+ mm Hg). This data has been used as an example of join exposure by Rothman for measuring interactions (see examples).

Usage

Rothman

Format

A labelled tibble with 477 rows and 3 variables:

stroke Thrombotic stroke, factor with levels "No" and "Yes".

oc Current user of oral contraceptives, factor with levels "Non-user" and "User".

ht Hypertension, factor with levels "No" (<160/95 mm Hg) and "Yes".

round_pval

Source

Collaborative Group for the Study of Stroke in Young Women (1975) Oral contraceptives and stroke in young women. JAMA 231:718-722.

Rothman, KJ (2002) Epidemiology. An Introduction. Oxford University Press.

Examples

```
data(Rothman)
mhor(stroke ~ ht / oc, data = Rothman)
## Model with standard interaction term:
model1 <- glm(stroke ~ ht * oc, data = Rothman, family = binomial)</pre>
glm_coef(model1)
## Model considering join exposure:
Rothman$join <- 0
Rothman$join[Rothman$oc == "Non-user" & Rothman$ht == "Yes"] <- 1
Rothman$join[Rothman$oc == "User" & Rothman$ht == "No"] <- 2</pre>
Rothman$join[Rothman$oc == "User" & Rothman$ht == "Yes"] <- 3</pre>
Rothman$join <- factor(Rothman$join, labels = c(</pre>
  "Unexposed", "Hypertension", "OC user",
  "OC and hypertension"
))
require(sjlabelled, quietly = TRUE)
Rothman$join <- set_label(Rothman$join, label = "Exposure")</pre>
model2 <- glm(stroke ~ join, data = Rothman, family = binomial)</pre>
glm_coef(model2)
```

round_pval

Rounding p-values.

Description

round_pval is an internal function called by glm_coef to round p-values from model coefficients.

Usage

```
round_pval(pval)
```

Arguments

pval vector of p-values, numeric.

Sandler

Description

A case-control study to investigate the effects of passive smoking on cancer. Passive smoking was defined as exposure to the cigarette smoke of a spouse who smoked at least one cigarette per day for at least 6 months.

Usage

Sandler

Format

A labelled tibble with 998 rows and 3 variables:

passive Passive smoker, factor with levels "No" and "Yes".

cancer Diagnosed with cancer, factor with levels "No" and "Yes".

smoke Active smoker, factor with levels "No" and "Yes".

Source

Sandler, DP, Everson, RB, Wilcox, AJ (1985). Passive smoking in adulthood and cancer risk. Amer J Epidem, 121: 37-48.

Examples

```
data(Sandler)
```

mhor(cancer ~ smoke / passive, data = Sandler)

Sharples

Measured and self-reported weight in New Zealand.

Description

Data on measured and self-reported weight from 40–50 year old participants in the 1989/1990 Life In New Zealand Survey.

Usage

Sharples

ss_jk

Format

A tibble with 343 rows and 4 variables:

srweight Self-reported weight in kg.

weight Measured weight in kg.

srbmi Body mass index calculated from self-reported weight and self-reported height in kg/m². **mbmi** Body mass index calculated from measured weight and measured height in kg/m².

Source

Sharples, H, et al. (2012) Agreement between measured and self-reported height, weight and BMI in predominantly European middle-aged New Zealanders: findings from a nationwide 1989 survey. New Zealand Med J 125: 60-69.

Examples

```
Sharples |>
    bland_altman(srweight ~ weight, transform = TRUE) |>
    gf_labs(x = "Mean of weights (kg)", y = "Measured weight / Self-reported weight")
```

ss_jk

Sum of squares for Jackknife.

Description

ss_jk is an internal function called by jack_knife. It calculates the squared difference of a numerical variable around a given value (for example, the mean).

Usage

ss_jk(obs, stat)

Arguments

obs	A numerical vector with no missing values (NA's).
stat	The value of the statistic that is used as a reference.

Value

The squared difference between a variable and a given value.

Examples

```
x <- rnorm(10, 170, 8)
x
mean(x)
ss_jk(x, mean(x))
jack_knife(x)</pre>
```

stats_quotes

Description

stats_quotes is an internal function called by estat.

Usage

```
stats_quotes(x, data2, digits = 2)
```

Arguments

х	a numeric variable
data2	A data frame where x can be found.
digits	Number of digits for rounding.

strip_error

Strip plots with error bars.

Description

strip_error constructs strip plots with error bars showing 95 confidence intervals around mean values.

Usage

```
strip_error(
   object = NULL,
   formula = NULL,
   data = NULL,
   pch = 20,
   size = 1,
   alpha = 0.7,
   col = "indianred3",
   ...
)
```

Arguments

object	When chaining, this holds an object produced in the earlier portions of the chain. Most users can safely ignore this argument. See details and examples.
formula	A formula with shape: $y \sim x$ or $y \sim x z$ where y is a numerical variable and both x and z are factors.

data	A data frame where the variables in the formula can be found.
pch	Point character passed to gf_point or gf_jitter.
size	Size of the symbol (pch) for representing data values.
alpha	Opacity of the symbol $(0 = invisible, 1 = opaque)$.
col	A colour or formula used for mapping colour.
	Additional information passed to gf_jitter or gf_point.

Examples

```
data(energy, package = "ISwR")
require(sjlabelled, quietly = TRUE)
energy <- energy |>
 var_labels(
   expend = "Energy expenditure (MJ/day)",
   stature = "Stature"
  )
energy |>
  strip_error(expend ~ stature, col = "red")
t.test(expend ~ stature, data = energy)
## Adding an horizontal line to show significant difference:
energy |>
  strip_error(expend ~ stature, col = "red") |>
  gf_star(1, 13, 2, 13.2, 13.4, "**")
data(birthwt, package = "MASS")
require(dplyr, quietly = TRUE)
birthwt <- birthwt |>
  mutate(
    smoke = factor(smoke, labels = c("Non-smoker", "Smoker")),
   Race = factor(race > 1, labels = c("White", "Non-white"))
  ) |>
  var_labels(
   bwt = "Birth weight (g)",
   smoke = "Smoking status"
  )
birthwt |>
  strip_error(bwt ~ smoke | Race, col = "darksalmon")
birthwt |>
  strip_error(bwt ~ smoke, col = ~Race)
birthwt |>
  strip_error(bwt ~ smoke, pch = ~Race, col = ~Race)
birthwt |>
  strip_error(bwt ~ smoke | Race)
```

Thall

Description

Randomised control trial of an antiepilectic drug (prograbide), in which the number of seizures of 59 patients at baseline and other four follow-up visits were recorded.

Usage

Thall

Format

A tibble with 59 rows and 8 variables:

id Subject ID.

treat Treatment, factor with levels "Control" and "Prograbide".

base Number of seizures at baseline.

- age Age in years at baseline.
- y1 Number of seizures at year one follow-up.
- y2 Number of seizures at year two follow-up.
- y3 Number of seizures at year three follow-up.
- y4 Number of seizures at year four follow-up.

Source

Thall, PF and Vail, SC (1990) Some covariance models for longitudinal count data with overdispersion. Biometrics, 46: 657-671.

Stukel, TA (1993) Comparison of methods for the analysis of longitudinal data. Statistics Med 12: 1339-1351.

Shoukri, MM and Chaudhary, MA (2007) Analysis of correlated data with SAS and R. Third Edition. Chapman & Hall/CRC.

Tuzson

Description

Data of peak knee velocity in walking at flexion and extension in studies about functional performance in cerebral palsy.

Usage

Tuzson

Format

A labelled tibble with 18 rows and 2 variables:

flexion Peak knee velocity in gait: flexion (degree/s).

extension Peak knee velocity in gait: extension (degree/s).

Source

Tuzson, AE, Granata, KP, and Abel, MF (2003) Spastic velocity threshold constrains functional performance in cerebral palsy. Arch Phys Med Rehabil 84: 1363-1368.

Examples

```
data(Tuzson)
Tuzson |>
  gf_point(flexion ~ extension)
cor.test(~ flexion + extension, data = Tuzson)
```

Vanderpump

Smoking and mortality in Whickham, England.

Description

Data represents women participating in a health survey in Whickham, England in 1972-1974.

Usage

Vanderpump

A labelled tibble with 1314 rows and 3 variables:

vstatus Vitality status, factor with levels "Alive" and "Death".

smoker Smoking status, factor with levels "Non-smoker" and "Smoker".

agegrp Age group, factor with levels "18-44", "45-64" and "64+".

Source

Vanderpump, MP, et al (1996) Thyroid, 6:155-160.

Appleton, DR, French, JM and Vanderpump, PJ (1996) Ignoring a covariate: An example of Simpson's paradox. The American Statistician 50:340-341.

Vittinghoff, E, Glidden, DV, Shiboski, SC and McCulloh, CE (2005) Regression methods in Biostatistics. Springer.

Examples

data(Vanderpump)

mhor(vstatus ~ agegrp / smoker, data = Vanderpump)

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