Package ‘insight’

August 10, 2022

Type Package

Title Easy Access to Model Information for Various Model Objects

Version 0.18.2

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Description A tool to provide an easy, intuitive and consistent access to information contained in various R models, like model formulas, model terms, information about random effects, data that was used to fit the model or data from response variables. 'insight' mainly revolves around two types of functions: Functions that find (the names of) information, starting with 'find_', and functions that get the underlying data, starting with 'get_'. The package has a consistent syntax and works with many different model objects, where otherwise functions to access these information are missing.

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URL https://easystats.github.io/insight/

BugReports https://github.com/easystats/insight/issues

Depends R (>= 3.5)

Imports methods, stats, utils

Suggests AER, afex, aod, BayesFactor, bayestestR, bbmle, bdsmatrix, betareg, bife, biglm, blavaan, blme, boot, brms, censReg, cgam, clubSandwich, coxme, cplm, crch, datawizard, effectsize, emmeans, epiR, estimatr, feisr, fixest, fungible, gam, gamlss, gamm4, gbm, gee, geepack, GLMMadaptive, glmmTMB, gmnl, gt, httr, ivreg, JM, knitr, lavaan, lfe, lme4, lmerTest, lmtest, logistf, MASS, marginaleffects, Matrix, mclust, MCMCglmm, merTools, metaBMA, mgcv, mice, mlogit, mhurdle, multgee, nlme, nnet, nonnest2, ordinal, panelr, parameters, parsnip, pbkrtest, performance, plm, poorman, pscl, psych, quantreg, rmarkdown, rms, robustbase, robustlmm, rstanarm (>= 2.21.1), rstantools, rstudioapi, sandwich, speedglm, splines, statmod, survey, survival, testthat, truncreg, tweedie, VGAM

VignetteBuilder knitr

Encoding UTF-8

Language en-US

RoxygenNote 7.2.1

Config/testthat/edition 3

NeedsCompilation no

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Repository CRAN

Date/Publication 2022-08-10 11:50:06 UTC

R topics documented:

`all_models_equal` *Checks if all objects are models of same class*

Description

Small helper that checks if all objects are *supported* (regression) model objects and of same class.

Usage

```
all_models_equal(..., verbose = FALSE)

all_models_same_class(..., verbose = FALSE)
```

Arguments

`...`	A list of objects.
`verbose`	Toggle off warnings.

Value

A logical, `TRUE` if `x` are all supported model objects of same class.

Examples

```
if (require("lme4")) {
  data(mtcars)
  data(sleepstudy)

  m1 <- lm(mpg ~ wt + cyl + vs, data = mtcars)
  m2 <- lm(mpg ~ wt + cyl, data = mtcars)
  m3 <- lmer(Reaction ~ Days + (1 | Subject), data = sleepstudy)
  m4 <- glm(formula = vs ~ wt, family = binomial(), data = mtcars)

  all_models_same_class(m1, m2)
  all_models_same_class(m1, m2, m3)
  all_models_same_class(m1, m4, m2, m3, verbose = TRUE)
  all_models_same_class(m1, m4, mtcars, m2, m3, verbose = TRUE)
}
```

`check_if_installed` *Checking if needed package is installed*

Description

Checking if needed package is installed

Usage

```
check_if_installed(
  package,
  reason = "for this function to work",
  stop = TRUE,
  minimum_version = NULL,
  quietly = FALSE,
  prompt = interactive(),
  ...
)
```

Arguments

`package`	A character vector naming the package(s), whose installation needs to be checked in any of the libraries.
`reason`	A phrase describing why the package is needed. The default is a generic description.
`stop`	Logical that decides whether the function should stop if the needed package is not installed.
`minimum_version`	A character vector, representing the minimum package version that is required for each package. Should be of same length as `package`. If `NULL`, no check for minimum version is done.
`quietly`	Logical, if `TRUE`, invisibly returns a vector of logicals (`TRUE` for each installed package, `FALSE` otherwise), and does not stop or throw a warning. If `quietly = TRUE`, arguments `stop` and `prompt` are ignored. Use this argument to internally check for package dependencies without stopping or warnings.
`prompt`	If `TRUE`, will prompt the user to install needed package(s). Ignored if `quietly = TRUE`.
`...`	Currently ignored

Value

If `stop = TRUE`, and `package` is not yet installed, the function stops and throws an error. Else, a named logical vector is returned, indicating which of the packages are installed, and which not.

Examples

```
## Not run:
check_if_installed("inexistent_package")
check_if_installed("insight")
check_if_installed("insight", minimum_version = "99.8.7")

x <- check_if_installed(c("inexistent", "also_not_here"), stop = FALSE)
x

## End(Not run)
```

`clean_names` *Get clean names of model terms*

Description

This function "cleans" names of model terms (or a character vector with such names) by removing patterns like `log()` or `as.factor()` etc.

Usage

```
clean_names(x, ...)

## S3 method for class 'character'
clean_names(x, include_names = FALSE, ...)
```

Arguments

`x`	A fitted model, or a character vector.
`...`	Currently not used.
`include_names`	Logical, if `TRUE`, returns a named vector where names are the original values of `x`.

Value

The "cleaned" variable names as character vector, i.e. pattern like `s()` for splines or `log()` are removed from the model terms.

Note

Typically, this method is intended to work on character vectors, in order to remove patterns that obscure the variable names. For convenience reasons it is also possible to call `clean_names()` also on a model object. If `x` is a regression model, this function is (almost) equal to calling `find_variables()`. The main difference is that `clean_names()` always returns a character vector, while `find_variables()` returns a list of character vectors, unless `flatten = TRUE`. See 'Examples'.

Examples

```
# example from ?stats::glm
counts <- c(18, 17, 15, 20, 10, 20, 25, 13, 12)
outcome <- as.numeric(gl(3, 1, 9))
treatment <- gl(3, 3)
m <- glm(counts ~ log(outcome) + as.factor(treatment), family = poisson())
clean_names(m)

# difference "clean_names()" and "find_variables()"
if (require("lme4")) {
  m <- glmer(
    cbind(incidence, size - incidence) ~ period + (1 | herd),
    data = cbpp,
    family = binomial
  )

  clean_names(m)
  find_variables(m)
  find_variables(m, flatten = TRUE)
}
```

clean_parameters *Get clean names of model parameters*

Description

This function "cleans" names of model parameters by removing patterns like `"r_"` or `"b[]"` (mostly applicable to Stan models) and adding columns with information to which group or component parameters belong (i.e. fixed or random, count or zero-inflated...)

The main purpose of this function is to easily filter and select model parameters, in particular of - but not limited to - posterior samples from Stan models, depending on certain characteristics. This might be useful when only selective results should be reported or results from all parameters should be filtered to return only certain results (see `print_parameters()`).

Usage

```
clean_parameters(x, ...)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.

Details

The `Effects` column indicate if a parameter is a *fixed* or *random* effect. The `Component` can either be *conditional* or *zero_inflated*. For models with random effects, the `Group` column indicates the grouping factor of the random effects. For multivariate response models from **brms** or **rstanarm**, an additional *Response* column is included, to indicate which parameters belong to which response formula. Furthermore, *Cleaned_Parameter* column is returned that contains "human readable" parameter names (which are mostly identical to `Parameter`, except for for models from **brms** or **rstanarm**, or for specific terms like smooth- or spline-terms).

Value

A data frame with "cleaned" parameter names and information on effects, component and group where parameters belong to. To be consistent across different models, the returned data frame always has at least four columns `Parameter`, `Effects`, `Component` and `Cleaned_Parameter`. See 'Details'.

Examples

```
## Not run:
library(brms)
model <- download_model("brms_zi_2")
clean_parameters(model)

## End(Not run)
```

color_if *Color-formatting for data columns based on condition*

Description

Convenient function that formats columns in data frames with color codes, where the color is chosen based on certain conditions. Columns are then printed in color in the console.

Usage

```
color_if(
  x,
  columns,
  predicate = `>`,
  value = 0,
  color_if = "green",
  color_else = "red",
  digits = 2
)

colour_if(
  x,
  columns,
  predicate = `>`,
  value = 0,
  colour_if = "green",
  colour_else = "red",
  digits = 2
)
```

Arguments

`x`	A data frame
`columns`	Character vector with column names of `x` that should be formatted.
`predicate`	A function that takes `columns` and `value` as input and which should return `TRUE` or `FALSE`, based on if the condition (in comparison with `value`) is met.
`value`	The comparator. May be used in conjunction with `predicate` to quickly set up a function which compares elements in `colums` to `value`. May be ignored when `predicate` is a function that internally computes other comparisons. See 'Examples'.
`color_if, colour_if`	Character vector, indicating the color code used to format values in `x` that meet the condition of `predicate` and `value`. May be one of `"red"`, `"yellow"`, `"green"`, `"blue"`, `"violet"`, `"cyan"` or `"grey"`. Formatting is also possible with `"bold"` or `"italic"`.

`color_else, colour_else`
See `color_if`, but only for conditions that are *not* met.

`digits` Digits for rounded values.

Details

The predicate-function simply works like this: `which(predicate(x[, columns], value))`

Value

The .

Examples

```
# all values in Sepal.Length larger than 5 in green, all remaining in red
x <- color_if(iris[1:10, ], columns = "Sepal.Length", predicate = `>`, value = 5)
x
cat(x$Sepal.Length)

# all levels "setosa" in Species in green, all remaining in red
x <- color_if(iris, columns = "Species", predicate = `==`, value = "setosa")
cat(x$Species)

# own function, argument "value" not needed here
p <- function(x, y) {
  x >= 4.9 & x <= 5.1
}
# all values in Sepal.Length between 4.9 and 5.1 in green, all remaining in red
x <- color_if(iris[1:10, ], columns = "Sepal.Length", predicate = p)
cat(x$Sepal.Length)
```

`compact_character` *Remove empty strings from character*

Description

Remove empty strings from character

Usage

```
compact_character(x)
```

Arguments

`x` A single character or a vector of characters.

Value

A character or a character vector with empty strings removed.

Examples

```
compact_character(c("x", "y", NA))
compact_character(c("x", "NULL", "", "y"))
```

---

compact_list *Remove empty elements from lists*

Description

Remove empty elements from lists

Usage

```
compact_list(x, remove_na = FALSE)
```

Arguments

x	A list or vector.
remove_na	Logical to decide if NAs should be removed.

Examples

```
compact_list(list(NULL, 1, c(NA, NA)))
compact_list(c(1, NA, NA))
compact_list(c(1, NA, NA), remove_na = TRUE)
```

---

display *Generic export of data frames into formatted tables*

Description

`display()` is a generic function to export data frames into various table formats (like plain text, markdown, ...). `print_md()` usually is a convenient wrapper for `display(format = "markdown")`. Similar, `print_html()` is a shortcut for `display(format = "html")`. See the documentation for the specific objects' classes.

Usage

```
display(object, ...)

print_md(x, ...)

print_html(x, ...)

## S3 method for class 'data.frame'
display(object, format = "markdown", ...)

## S3 method for class 'data.frame'
print_md(x, ...)

## S3 method for class 'data.frame'
print_html(x, ...)
```

Arguments

`object, x`	A data frame.
`...`	Arguments passed to other methods.
`format`	String, indicating the output format. Can be `"markdown"` or `"html"`.

Value

Depending on `format`, either an object of class `gt_tbl` or a character vector of class `knitr_kable`.

Examples

```
display(iris[1:5, ])
```

---

download_model   *Download circus models*

Description

Downloads pre-compiled models from the *circus*-repository. The *circus*-repository contains a variety of fitted models to help the systematic testing of other packages

Usage

```
download_model(name, url = NULL)
```

Arguments

`name`	Model name.
`url`	String with the URL from where to download the model data. Optional, and should only be used in case the repository-URL is changing. By default, models are downloaded from `https://raw.github.com/easystats/circus/master/data/`.

Details

The code that generated the model is available at the https://easystats.github.io/circus/reference/index.html.

Value

A model from the *circus*-repository.

References

https://easystats.github.io/circus/

`ellipsis_info` *Gather information about objects in ellipsis (dot dot dot)*

Description

Provides information regarding the models entered in an ellipsis. It detects whether all are models, regressions, nested regressions etc., assigning different classes to the list of objects.

Usage

```
ellipsis_info(objects, ...)

## Default S3 method:
ellipsis_info(..., only_models = TRUE, verbose = TRUE)
```

Arguments

`objects, ...`	Arbitrary number of objects. May also be a list of model objects.
`only_models`	Only keep supported models (default to `TRUE`).
`verbose`	Toggle warnings.

Value

The list with objects that were passed to the function, including additional information as attributes (e.g. if models have same response or are nested).

Examples

```
m1 <- lm(Sepal.Length ~ Petal.Width + Species, data = iris)
m2 <- lm(Sepal.Length ~ Species, data = iris)
m3 <- lm(Sepal.Length ~ Petal.Width, data = iris)
m4 <- lm(Sepal.Length ~ 1, data = iris)
m5 <- lm(Petal.Width ~ 1, data = iris)

objects <- ellipsis_info(m1, m2, m3, m4)
```

```
class(objects)

objects <- ellipsis_info(m1, m2, m4)
attributes(objects)$is_nested

objects <- ellipsis_info(m1, m2, m5)
attributes(objects)$same_response
```

---

export_table *Data frame and Tables Pretty Formatting*

Description

Data frame and Tables Pretty Formatting

Usage

```
export_table(
  x,
  sep = " | ",
  header = "-",
  cross = NULL,
  empty_line = NULL,
  digits = 2,
  protect_integers = TRUE,
  missing = "",
  width = NULL,
  format = NULL,
  title = NULL,
  caption = title,
  subtitle = NULL,
  footer = NULL,
  align = NULL,
  group_by = NULL,
  zap_small = FALSE,
  table_width = NULL,
  verbose = TRUE
)
```

Arguments

x	A data frame. May also be a list of data frames, to export multiple data frames into multiple tables.
sep	Column separator.
header	Header separator. Can be NULL.
cross	Character that is used where separator and header lines cross.

empty_line
: Separator used for empty lines. If NULL, line remains empty (i.e. filled with whitespaces).

digits
: Number of digits for rounding or significant figures. May also be "signif" to return significant figures or "scientific" to return scientific notation. Control the number of digits by adding the value as suffix, e.g. digits = "scientific4" to have scientific notation with 4 decimal places, or digits = "signif5" for 5 significant figures (see also signif()).

protect_integers
: Should integers be kept as integers (i.e., without decimals)?

missing
: Value by which NA values are replaced. By default, an empty string (i.e. "") is returned for NA.

width
: Refers to the width of columns (with numeric values). Can be either NULL, a number or a named numeric vector. If NULL, the width for each column is adjusted to the minimum required width. If a number, columns with numeric values will have the minimum width specified in width. If a named numeric vector, value names are matched against column names, and for each match, the specified width is used (see 'Examples'). Only applies to text-format (see format).

format
: Name of output-format, as string. If NULL (or "text"), returned output is used for basic printing. Can be one of NULL (the default) resp. "text" for plain text, "markdown" (or "md") for markdown and "html" for HTML output.

title, caption, subtitle
: Table title (same as caption) and subtitle, as strings. If NULL, no title or subtitle is printed, unless it is stored as attributes (table_title, or its alias table_caption, and table_subtitle). If x is a list of data frames, caption may be a list of table captions, one for each table.

footer
: Table footer, as string. For markdown-formatted tables, table footers, due to the limitation in markdown rendering, are actually just a new text line under the table. If x is a list of data frames, footer may be a list of table captions, one for each table.

align
: Column alignment. For markdown-formatted tables, the default align = NULL will right-align numeric columns, while all other columns will be left-aligned. If format = "html", the default is left-align first column and center all remaining. May be a string to indicate alignment rules for the complete table, like "left", "right", "center" or "firstleft" (to left-align first column, center remaining); or maybe a string with abbreviated alignment characters, where the length of the string must equal the number of columns, for instance, align = "lccrl" would left-align the first column, center the second and third, right-align column four and left-align the fifth column. For HTML-tables, may be one of "center", "left" or "right".

group_by
: Name of column in x that indicates grouping for tables. Only applies when format = "html". group_by is passed down to gt::gt(groupname_col = group_by).

zap_small
: Logical, if TRUE, small values are rounded after digits decimal places. If FALSE, values with more decimal places than digits are printed in scientific notation.

table_width
: Numeric, or "auto", indicating the width of the complete table. If table_width = "auto" and the table is wider than the current width (i.e. line length) of the console (or any other source for textual output, like markdown files), the table is split into two parts. Else, if table_width is numeric and table rows are larger than table_width, the table is split into two parts.

verbose
: Toggle messages and warnings.

Value

A data frame in character format.

Note

The values for caption, subtitle and footer can also be provided as attributes of x, e.g. if caption = NULL and x has attribute table_caption, the value for this attribute will be used as table caption. table_subtitle is the attribute for subtitle, and table_footer for footer.

See Also

Vignettes Formatting, printing and exporting tables and Formatting model parameters.

Examples

```
export_table(head(iris))
export_table(head(iris), cross = "+")
export_table(head(iris), sep = " ", header = "*", digits = 1)

# split longer tables
export_table(head(iris), table_width = 30)

## Not run:
# colored footers
data(iris)
x <- as.data.frame(iris[1:5, ])
attr(x, "table_footer") <- c("This is a yellow footer line.", "yellow")
export_table(x)

attr(x, "table_footer") <- list(
  c("\nA yellow line", "yellow"),
  c("\nAnd a red line", "red"),
  c("\nAnd a blue line", "blue")
)
export_table(x)

attr(x, "table_footer") <- list(
  c("Without the ", "yellow"),
  c("new-line character ", "red"),
  c("we can have multiple colors per line.", "blue")
)
export_table(x)
```

```
## End(Not run)

# column-width
d <- data.frame(
  x = c(1, 2, 3),
  y = c(100, 200, 300),
  z = c(10000, 20000, 30000)
)
export_table(d)
export_table(d, width = 8)
export_table(d, width = c(x = 5, z = 10))
export_table(d, width = c(x = 5, y = 5, z = 10), align = "lcr")
```

`find_algorithm` *Find sampling algorithm and optimizers*

Description

Returns information on the sampling or estimation algorithm as well as optimization functions, or for Bayesian model information on chains, iterations and warmup-samples.

Usage

```
find_algorithm(x, ...)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.

Value

A list with elements depending on the model.
For frequentist models:

- `algorithm`, for instance `"OLS"` or `"ML"`
- `optimizer`, name of optimizing function, only applies to specific models (like `gam`)

For frequentist mixed models:

- `algorithm`, for instance `"REML"` or `"ML"`
- `optimizer`, name of optimizing function

For Bayesian models:

- `algorithm`, the algorithm
- `chains`, number of chains
- `iterations`, number of iterations per chain
- `warmup`, number of warmups per chain

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  m <- lmer(Reaction ~ Days + (1 | Subject), data = sleepstudy)
  find_algorithm(m)
}
## Not run:
library(rstanarm)
m <- stan_lmer(Reaction ~ Days + (1 | Subject), data = sleepstudy)
find_algorithm(m)

## End(Not run)
```

---

find_formula *Find model formula*

Description

Returns the formula(s) for the different parts of a model (like fixed or random effects, zero-inflated component, ...). `formula_ok()` checks if a model formula has valid syntax regarding writing `TRUE` instead of `T` inside `poly()` and that no data names are used (i.e. no `data$variable`, but rather `variable`).

Usage

```
find_formula(x, verbose = TRUE, ...)

formula_ok(x, verbose = TRUE, ...)
```

Arguments

`x`	A fitted model.
`verbose`	Toggle warnings.
`...`	Currently not used.

Value

A list of formulas that describe the model. For simple models, only one list-element, `conditional`, is returned. For more complex models, the returned list may have following elements:

- `conditional`, the "fixed effects" part from the model (in the context of fixed-effects or instrumental variable regression, also called *regressors*) . One exception are `DirichletRegModel` models from **DirichletReg**, which has two or three components, depending on `model`.
- `random`, the "random effects" part from the model (or the `id` for gee-models and similar)
- `zero_inflated`, the "fixed effects" part from the zero-inflation component of the model
- `zero_inflated_random`, the "random effects" part from the zero-inflation component of the model

- `dispersion`, the dispersion formula
- `instruments`, for fixed-effects or instrumental variable regressions like `ivreg::ivreg()`, `lfe::felm()` or `plm::plm()`, the instrumental variables
- `cluster`, for fixed-effects regressions like `lfe::felm()`, the cluster specification
- `correlation`, for models with correlation-component like `nlme::gls()`, the formula that describes the correlation structure
- `slopes`, for fixed-effects individual-slope models like `feisr::feis()`, the formula for the slope parameters
- `precision`, for `DirichletRegModel` models from **DirichletReg**, when parametrization (i.e. `model`) is `"alternative"`.

Note

For models of class `lme` or `gls` the correlation-component is only returned, when it is explicitly defined as named argument (`form`), e.g. `corAR1(form = ~1 | Mare)`

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_formula(m)

if (require("lme4")) {
  m <- lmer(Sepal.Length ~ Sepal.Width + (1 | Species), data = iris)
  f <- find_formula(m)
  f
  format(f)
}
```

`find_interactions` *Find interaction terms from models*

Description

Returns all lowest to highest order interaction terms from a model.

Usage

```
find_interactions(
  x,
  component = c("all", "conditional", "zi", "zero_inflated", "dispersion", "instruments"),
  flatten = FALSE
)
```

Arguments

`x`	A fitted model.
`component`	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.

Value

A list of character vectors that represent the interaction terms. Depending on `component`, the returned list has following elements (or `NULL`, if model has no interaction term):

- `conditional`, interaction terms that belong to the "fixed effects" terms from the model
- `zero_inflated`, interaction terms that belong to the "fixed effects" terms from the zero-inflation component of the model
- `instruments`, for fixed-effects regressions like `ivreg`, `felm` or `plm`, interaction terms that belong to the instrumental variables

Examples

```
data(mtcars)

m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_interactions(m)

m <- lm(mpg ~ wt * cyl + vs * hp * gear + carb, data = mtcars)
find_interactions(m)
```

`find_offset` *Find possible offset terms in a model*

Description

Returns a character vector with the name(s) of offset terms.

Usage

```
find_offset(x)
```

Arguments

`x`	A fitted model.

Value

A character vector with the name(s) of offset terms.

Examples

```
# Generate some zero-inflated data
set.seed(123)
N <- 100 # Samples
x <- runif(N, 0, 10) # Predictor
off <- rgamma(N, 3, 2) # Offset variable
yhat <- -1 + x * 0.5 + log(off) # Prediction on log scale
dat <- data.frame(y = NA, x, logOff = log(off))
dat$y <- rpois(N, exp(yhat)) # Poisson process
dat$y <- ifelse(rbinom(N, 1, 0.3), 0, dat$y) # Zero-inflation process

if (require("pscl")) {
  m1 <- zeroinfl(y ~ offset(logOff) + x | 1, data = dat, dist = "poisson")
  find_offset(m1)

  m2 <- zeroinfl(y ~ x | 1, data = dat, offset = logOff, dist = "poisson")
  find_offset(m2)
}
```

`find_parameters` *Find names of model parameters*

Description

Returns the names of model parameters, like they typically appear in the `summary()` output. For Bayesian models, the parameter names equal the column names of the posterior samples after coercion from `as.data.frame()`. See the documentation for your object's class:

- Bayesian models (**rstanarm**, **brms**, **MCMCglmm**, ...)
- Generalized additive models (**mgcv**, **VGAM**, ...)
- Marginal effects models (**mfx**)
- Estimated marginal means (**emmeans**)
- Mixed models (**lme4**, **glmmTMB**, **GLMMadaptive**, ...)
- Zero-inflated and hurdle models (**pscl**, ...)
- Models with special components (**betareg**, **MuMIn**, ...)

Usage

```
find_parameters(x, ...)

## Default S3 method:
find_parameters(x, flatten = FALSE, verbose = TRUE, ...)
```

Arguments

x	A fitted model.
...	Currently not used.
flatten	Logical, if TRUE, the values are returned as character vector, not as list. Duplicated values are removed.
verbose	Toggle messages and warnings.

Value

A list of parameter names. For simple models, only one list-element, conditional, is returned.

Model components

Possible values for the component argument depend on the model class. Following are valid options:

- "all": returns all model components, applies to all models, but will only have an effect for models with more than just the conditional model component.
- "conditional": only returns the conditional component, i.e. "fixed effects" terms from the model. Will only have an effect for models with more than just the conditional model component.
- "smooth_terms": returns smooth terms, only applies to GAMs (or similar models that may contain smooth terms).
- "zero_inflated" (or "zi"): returns the zero-inflation component.
- "dispersion": returns the dispersion model component. This is common for models with zero-inflation or that can model the dispersion parameter.
- "instruments": for instrumental-variable or some fixed effects regression, returns the instruments.
- "location": returns location parameters such as conditional, zero_inflated, smooth_terms, or instruments (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters).
- "distributional" (or "auxiliary"): components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

find_parameters.averaging
Find model parameters from models with special components

Description

Returns the names of model parameters, like they typically appear in the summary() output.

Usage

```
## S3 method for class 'averaging'
find_parameters(x, component = c("conditional", "full"), flatten = FALSE, ...)

## S3 method for class 'betareg'
find_parameters(
  x,
  component = c("all", "conditional", "precision", "location", "distributional",
    "auxiliary"),
  flatten = FALSE,
  ...
)

## S3 method for class 'DirichletRegModel'
find_parameters(
  x,
  component = c("all", "conditional", "precision", "location", "distributional",
    "auxiliary"),
  flatten = FALSE,
  ...
)

## S3 method for class 'mjoint'
find_parameters(
  x,
  component = c("all", "conditional", "survival"),
  flatten = FALSE,
  ...
)

## S3 method for class 'glmx'
find_parameters(
  x,
  component = c("all", "conditional", "extra"),
  flatten = FALSE,
  ...
)
```

Arguments

x
: A fitted model.

component
: Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional`, `zero_inflated`, `smooth_terms`, or `instruments` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma`, `dispersion`, `beta` or `precision` (and other auxiliary parameters) are returned.

flatten
: Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.

...
: Currently not used.

Value

A list of parameter names. The returned list may have following elements:

- `conditional`, the "fixed effects" part from the model.
- `full`, parameters from the full model.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

---

find_parameters.betamfx

Find names of model parameters from marginal effects models

Description

Returns the names of model parameters, like they typically appear in the `summary()` output.

Usage

```
## S3 method for class 'betamfx'
find_parameters(
  x,
  component = c("all", "conditional", "precision", "marginal", "location",
```

```
    "distributional", "auxiliary"),
  flatten = FALSE,
  ...
)

## S3 method for class 'logitmfx'
find_parameters(
  x,
  component = c("all", "conditional", "marginal", "location"),
  flatten = FALSE,
  ...
)
```

Arguments

`x`	A fitted model.
`component`	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional`, `zero_inflated`, `smooth_terms`, or `instruments` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma`, `dispersion`, `beta` or `precision` (and other auxiliary parameters) are returned.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
`...`	Currently not used.

Value

A list of parameter names. The returned list may have following elements:

- `conditional`, the "fixed effects" part from the model.
- `marginal`, the marginal effects.
- `precision`, the precision parameter.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

find_parameters.BGGM *Find names of model parameters from Bayesian models*

Description

Returns the names of model parameters, like they typically appear in the `summary()` output. For Bayesian models, the parameter names equal the column names of the posterior samples after coercion from `as.data.frame()`.

Usage

```
## S3 method for class 'BGGM'
find_parameters(
  x,
  component = c("correlation", "conditional", "intercept", "all"),
  flatten = FALSE,
  ...
)

## S3 method for class 'BFBayesFactor'
find_parameters(
  x,
  effects = c("all", "fixed", "random"),
  component = c("all", "extra"),
  flatten = FALSE,
  ...
)

## S3 method for class 'MCMCglmm'
find_parameters(x, effects = c("all", "fixed", "random"), flatten = FALSE, ...)

## S3 method for class 'bamlss'
find_parameters(
  x,
  flatten = FALSE,
  component = c("all", "conditional", "location", "distributional", "auxiliary"),
  parameters = NULL,
  ...
)

## S3 method for class 'brmsfit'
find_parameters(
  x,
  effects = "all",
  component = "all",
  flatten = FALSE,
  parameters = NULL,
```

```
  ...
)

## S3 method for class 'bayesx'
find_parameters(
  x,
  component = c("all", "conditional", "smooth_terms"),
  flatten = FALSE,
  parameters = NULL,
  ...
)

## S3 method for class 'stanreg'
find_parameters(
  x,
  effects = c("all", "fixed", "random"),
  component = c("location", "all", "conditional", "smooth_terms", "sigma",
    "distributional", "auxiliary"),
  flatten = FALSE,
  parameters = NULL,
  ...
)

## S3 method for class 'sim.merMod'
find_parameters(
  x,
  effects = c("all", "fixed", "random"),
  flatten = FALSE,
  parameters = NULL,
  ...
)
```

Arguments

`x` A fitted model.

`component` Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional`, `zero_inflated`, `smooth_terms`, or `instruments` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma`, `dispersion`, `beta` or `precision` (and other auxiliary parameters) are returned.

`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
`...`	Currently not used.
`effects`	Should parameters for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`parameters`	Regular expression pattern that describes the parameters that should be returned.

Value

A list of parameter names. For simple models, only one list-element, `conditional`, is returned. For more complex models, the returned list may have following elements:

- `conditional`, the "fixed effects" part from the model
- `random`, the "random effects" part from the model
- `zero_inflated`, the "fixed effects" part from the zero-inflation component of the model
- `zero_inflated_random`, the "random effects" part from the zero-inflation component of the model
- `smooth_terms`, the smooth parameters

Furthermore, some models, especially from **brms**, can also return auxiliary parameters. These may be one of the following:

- `sigma`, the residual standard deviation (auxiliary parameter)
- `dispersion`, the dispersion parameters (auxiliary parameter)
- `beta`, the beta parameter (auxiliary parameter)
- `simplex`, simplex parameters of monotonic effects (**brms** only)
- `mix`, mixture parameters (**brms** only)
- `shiftprop`, shifted proportion parameters (**brms** only)

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

`find_parameters.emmGrid`

Find model parameters from estimated marginal means objects

Description

Returns the parameter names from a model.

Usage

```
## S3 method for class 'emmGrid'
find_parameters(x, flatten = FALSE, merge_parameters = FALSE, ...)
```

Arguments

`x`	A fitted model.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
`merge_parameters`	Logical, if `TRUE` and `x` has multiple columns for parameter names (like `emmGrid` objects may have), these are merged into a single parameter column, with parameters names and values as values.
`...`	Currently not used.

Value

A list of parameter names. For simple models, only one list-element, `conditional`, is returned.

Examples

```
data(mtcars)
model <- lm(mpg ~ wt * factor(cyl), data = mtcars)
if (require("emmeans", quietly = TRUE)) {
  emm <- emmeans(model, c("wt", "cyl"))
  find_parameters(emm)
}
```

---

find_parameters.gamlss

Find names of model parameters from generalized additive models

---

Description

Returns the names of model parameters, like they typically appear in the `summary()` output.

Usage

```
## S3 method for class 'gamlss'
find_parameters(x, flatten = FALSE, ...)

## S3 method for class 'gam'
find_parameters(
  x,
  component = c("all", "conditional", "smooth_terms", "location"),
  flatten = FALSE,
  ...
)
```

Arguments

x	A fitted model.
flatten	Logical, if TRUE, the values are returned as character vector, not as list. Duplicated values are removed.
...	Currently not used.
component	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: component = "all" returns all possible parameters. If component = "location", location parameters such as conditional, zero_inflated, smooth_terms, or instruments are returned (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters). For component = "distributional" (or "auxiliary"), components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.

Value

A list of parameter names. The returned list may have following elements:

- conditional, the "fixed effects" part from the model.
- smooth_terms, the smooth parameters.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

---

find_parameters.glmmTMB

Find names of model parameters from mixed models

Description

Returns the names of model parameters, like they typically appear in the summary() output.

Usage

```
## S3 method for class 'glmmTMB'
find_parameters(
  x,
  effects = c("all", "fixed", "random"),
```

```
  component = c("all", "conditional", "zi", "zero_inflated", "dispersion"),
  flatten = FALSE,
  ...
)

## S3 method for class 'merMod'
find_parameters(x, effects = c("all", "fixed", "random"), flatten = FALSE, ...)
```

Arguments

`x`	A fitted model.
`effects`	Should parameters for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`component`	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model or the dispersion term? Applies to models with zero-inflated and/or dispersion formula. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional` or `zero_inflated` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma` or `dispersion` (and other auxiliary parameters) are returned.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
`...`	Currently not used.

Value

A list of parameter names. The returned list may have following elements:

- `conditional`, the "fixed effects" part from the model.
- `random`, the "random effects" part from the model.
- `zero_inflated`, the "fixed effects" part from the zero-inflation component of the model.
- `zero_inflated_random`, the "random effects" part from the zero-inflation component of the model.
- `dispersion`, the dispersion parameters (auxiliary parameter)

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

find_parameters.zeroinfl

Find names of model parameters from zero-inflated models

Description

Returns the names of model parameters, like they typically appear in the summary() output.

Usage

```
## S3 method for class 'zeroinfl'
find_parameters(
  x,
  component = c("all", "conditional", "zi", "zero_inflated"),
  flatten = FALSE,
  ...
)

## S3 method for class 'mhurdle'
find_parameters(
  x,
 component = c("all", "conditional", "zi", "zero_inflated", "infrequent_purchase", "ip",
    "auxiliary"),
  flatten = FALSE,
  ...
)
```

Arguments

x	A fitted model.
component	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: component = "all" returns all possible parameters. If component = "location", location parameters such as conditional, zero_inflated, smooth_terms, or instruments are returned (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters). For component = "distributional" (or "auxiliary"), components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.
flatten	Logical, if TRUE, the values are returned as character vector, not as list. Duplicated values are removed.
...	Currently not used.

Value

A list of parameter names. The returned list may have following elements:

- `conditional`, the "fixed effects" part from the model.
- `zero_inflated`, the "fixed effects" part from the zero-inflation component of the model.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_parameters(m)
```

`find_predictors` *Find names of model predictors*

Description

Returns the names of the predictor variables for the different parts of a model (like fixed or random effects, zero-inflated component, ...). Unlike `find_parameters()`, the names from `find_predictors()` match the original variable names from the data that was used to fit the model.

Usage

```
find_predictors(x, ...)

## Default S3 method:
find_predictors(
  x,
  effects = c("fixed", "random", "all"),
 component = c("all", "conditional", "zi", "zero_inflated", "dispersion", "instruments",
    "correlation", "smooth_terms"),
  flatten = FALSE,
  verbose = TRUE,
  ...
)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.
`effects`	Should variables for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`component`	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.

`flatten` Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.

`verbose` Toggle warnings.

Value

A list of character vectors that represent the name(s) of the predictor variables. Depending on the combination of the arguments `effects` and `component`, the returned list has following elements:

- `conditional`, the "fixed effects" terms from the model
- `random`, the "random effects" terms from the model
- `zero_inflated`, the "fixed effects" terms from the zero-inflation component of the model
- `zero_inflated_random`, the "random effects" terms from the zero-inflation component of the model
- `dispersion`, the dispersion terms
- `instruments`, for fixed-effects regressions like `ivreg`, `felm` or `plm`, the instrumental variables
- `correlation`, for models with correlation-component like `gls`, the variables used to describe the correlation structure

Model components

Possible values for the `component` argument depend on the model class. Following are valid options:

- `"all"`: returns all model components, applies to all models, but will only have an effect for models with more than just the conditional model component.
- `"conditional"`: only returns the conditional component, i.e. "fixed effects" terms from the model. Will only have an effect for models with more than just the conditional model component.
- `"smooth_terms"`: returns smooth terms, only applies to GAMs (or similar models that may contain smooth terms).
- `"zero_inflated"` (or `"zi"`): returns the zero-inflation component.
- `"dispersion"`: returns the dispersion model component. This is common for models with zero-inflation or that can model the dispersion parameter.
- `"instruments"`: for instrumental-variable or some fixed effects regression, returns the instruments.
- `"location"`: returns location parameters such as `conditional`, `zero_inflated`, `smooth_terms`, or `instruments` (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters).
- `"distributional"` (or `"auxiliary"`): components like `sigma`, `dispersion`, `beta` or `precision` (and other auxiliary parameters) are returned.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_predictors(m)
```

`find_random` *Find names of random effects*

Description

Return the name of the grouping factors from mixed effects models.

Usage

```
find_random(x, split_nested = FALSE, flatten = FALSE)
```

Arguments

`x`	A fitted mixed model.
`split_nested`	Logical, if `TRUE`, terms from nested random effects will be returned as separated elements, not as single string with colon. See 'Examples'.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.

Value

A list of character vectors that represent the name(s) of the random effects (grouping factors). Depending on the model, the returned list has following elements:

- `random`, the "random effects" terms from the conditional part of model
- `zero_inflated_random`, the "random effects" terms from the zero-inflation component of the model

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  sleepstudy$mygrp <- sample(1:5, size = 180, replace = TRUE)
  sleepstudy$mysubgrp <- NA
  for (i in 1:5) {
    filter_group <- sleepstudy$mygrp == i
    sleepstudy$mysubgrp[filter_group] <-
      sample(1:30, size = sum(filter_group), replace = TRUE)
  }

  m <- lmer(
    Reaction ~ Days + (1 | mygrp / mysubgrp) + (1 | Subject),
    data = sleepstudy
  )

  find_random(m)
  find_random(m, split_nested = TRUE)
}
```

find_random_slopes *Find names of random slopes*

Description

Return the name of the random slopes from mixed effects models.

Usage

```
find_random_slopes(x)
```

Arguments

x	A fitted mixed model.

Value

A list of character vectors with the name(s) of the random slopes, or `NULL` if model has no random slopes. Depending on the model, the returned list has following elements:

- `random`, the random slopes from the conditional part of model
- `zero_inflated_random`, the random slopes from the zero-inflation component of the model

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  m <- lmer(Reaction ~ Days + (1 + Days | Subject), data = sleepstudy)
  find_random_slopes(m)
}
```

find_response *Find name of the response variable*

Description

Returns the name(s) of the response variable(s) from a model object.

Usage

```
find_response(x, combine = TRUE, ...)
```

Arguments

`x`	A fitted model.
`combine`	Logical, if `TRUE` and the response is a matrix-column, the name of the response matches the notation in formula, and would for instance also contain patterns like `"cbind(...)"`. Else, the original variable names from the matrix-column are returned. See 'Examples'.
`...`	Currently not used.

Value

The name(s) of the response variable(s) from `x` as character vector, or `NULL` if response variable could not be found.

Examples

```
if (require("lme4")) {
  data(cbpp)
  cbpp$trials <- cbpp$size - cbpp$incidence
  m <- glm(cbind(incidence, trials) ~ period, data = cbpp, family = binomial)

  find_response(m, combine = TRUE)
  find_response(m, combine = FALSE)
}
```

`find_smooth` *Find smooth terms from a model object*

Description

Return the names of smooth terms from a model object.

Usage

```
find_smooth(x, flatten = FALSE)
```

Arguments

`x`	A (gam) model.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.

Value

A character vector with the name(s) of the smooth terms.

Examples

```
if (require("mgcv")) {
  data(iris)
  model <- gam(Petal.Length ~ Petal.Width + s(Sepal.Length), data = iris)
  find_smooth(model)
}
```

`find_statistic` *Find statistic for model*

Description

Returns the statistic for a regression model (*t*-statistic, *z*-statistic, etc.).

Small helper that checks if a model is a regression model object and return the statistic used.

Usage

```
find_statistic(x, ...)
```

Arguments

`x`	An object.
`...`	Currently not used.

Value

A character describing the type of statistic. If there is no statistic available with a distribution, `NULL` will be returned.

Examples

```
# regression model object
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
find_statistic(m)
```

`find_terms` *Find all model terms*

Description

Returns a list with the names of all terms, including response value and random effects, "as is". This means, on-the-fly tranformations or arithmetic expressions like `log()`, `I()`, `as.factor()` etc. are preserved.

Usage

```
find_terms(x, flatten = FALSE, verbose = TRUE, ...)
```

Arguments

`x`	A fitted model.
`flatten`	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
`verbose`	Toggle warnings.
`...`	Currently not used.

Value

A list with (depending on the model) following elements (character vectors):

- `response`, the name of the response variable
- `conditional`, the names of the predictor variables from the *conditional* model (as opposed to the zero-inflated part of a model)
- `random`, the names of the random effects (grouping factors)
- `zero_inflated`, the names of the predictor variables from the *zero-inflated* part of the model
- `zero_inflated_random`, the names of the random effects (grouping factors)
- `dispersion`, the name of the dispersion terms
- `instruments`, the names of instrumental variables

Returns `NULL` if no terms could be found (for instance, due to problems in accessing the formula).

Note

The difference to `find_variables()` is that `find_terms()` may return a variable multiple times in case of multiple transformations (see examples below), while `find_variables()` returns each variable name only once.

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  m <- lmer(
    log(Reaction) ~ Days + I(Days^2) + (1 + Days + exp(Days) | Subject),
    data = sleepstudy
  )

  find_terms(m)
}
```

`find_transformation` *Find possible transformation of response variables*

Description

This functions checks whether any transformation, such as log- or exp-transforming, was applied to the response variable (dependent variable) in a regression formula. Currently, following patterns are detected: `log`, `log1p`, `log2`, `log10`, `exp`, `expm1`, `sqrt`, `log(x+<number>)`, `log-log` and `power` (to 2nd power, like `I(x^2)`).

Usage

```
find_transformation(x)
```

Arguments

`x`	A regression model.

Value

A string, with the name of the function of the applied transformation. Returns `"identity"` for no transformation, and e.g. `"log(x+3)"` when a specific values was added to the response variables before log-transforming. For unknown transformations, returns `NULL`.

Examples

```
# identity, no transformation
model <- lm(Sepal.Length ~ Species, data = iris)
find_transformation(model)

# log-transformation
model <- lm(log(Sepal.Length) ~ Species, data = iris)
find_transformation(model)

# log+2
model <- lm(log(Sepal.Length + 2) ~ Species, data = iris)
find_transformation(model)
```

find_variables *Find names of all variables*

Description

Returns a list with the names of all variables, including response value and random effects.

Usage

```
find_variables(
  x,
  effects = "all",
  component = "all",
  flatten = FALSE,
  verbose = TRUE
)
```

Arguments

x	A fitted model.
effects	Should variables for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
component	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
flatten	Logical, if `TRUE`, the values are returned as character vector, not as list. Duplicated values are removed.
verbose	Toggle warnings.

Value

A list with (depending on the model) following elements (character vectors):

- `response`, the name of the response variable
- `conditional`, the names of the predictor variables from the *conditional* model (as opposed to the zero-inflated part of a model)
- `cluster`, the names of cluster or grouping variables
- `dispersion`, the name of the dispersion terms
- `instruments`, the names of instrumental variables
- `random`, the names of the random effects (grouping factors)
- `zero_inflated`, the names of the predictor variables from the *zero-inflated* part of the model
- `zero_inflated_random`, the names of the random effects (grouping factors)

Model components

Possible values for the component argument depend on the model class. Following are valid options:

- "all": returns all model components, applies to all models, but will only have an effect for models with more than just the conditional model component.
- "conditional": only returns the conditional component, i.e. "fixed effects" terms from the model. Will only have an effect for models with more than just the conditional model component.
- "smooth_terms": returns smooth terms, only applies to GAMs (or similar models that may contain smooth terms).
- "zero_inflated" (or "zi"): returns the zero-inflation component.
- "dispersion": returns the dispersion model component. This is common for models with zero-inflation or that can model the dispersion parameter.
- "instruments": for instrumental-variable or some fixed effects regression, returns the instruments.
- "location": returns location parameters such as conditional, zero_inflated, smooth_terms, or instruments (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters).
- "distributional" (or "auxiliary"): components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.

Note

The difference to find_terms() is that find_variables() returns each variable name only once, while find_terms() may return a variable multiple times in case of transformations or when arithmetic expressions were used in the formula.

Examples

```
if (require("lme4")) {
  data(cbpp)
  data(sleepstudy)
  # some data preparation...
  cbpp$trials <- cbpp$size - cbpp$incidence
  sleepstudy$mygrp <- sample(1:5, size = 180, replace = TRUE)
  sleepstudy$mysubgrp <- NA
  for (i in 1:5) {
    filter_group <- sleepstudy$mygrp == i
    sleepstudy$mysubgrp[filter_group] <-
      sample(1:30, size = sum(filter_group), replace = TRUE)
  }

  m1 <- glmer(
    cbind(incidence, size - incidence) ~ period + (1 | herd),
    data = cbpp,
    family = binomial
  )
```

```
  find_variables(m1)

  m2 <- lmer(
    Reaction ~ Days + (1 | mygrp / mysubgrp) + (1 | Subject),
    data = sleepstudy
  )
  find_variables(m2)
  find_variables(m2, flatten = TRUE)
}
```

find_weights   *Find names of model weights*

Description

Returns the name of the variable that describes the weights of a model.

Usage

```
find_weights(x, ...)
```

Arguments

x	A fitted model.
...	Currently not used.

Value

The name of the weighting variable as character vector, or NULL if no weights were specified.

Examples

```
data(mtcars)
mtcars$weight <- rnorm(nrow(mtcars), 1, .3)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars, weights = weight)
find_weights(m)
```

fish   *Sample data set*

Description

A sample data set, used in tests and some examples.

format_bf *Bayes Factor formatting*

Description

Bayes Factor formatting

Usage

```
format_bf(
  bf,
  stars = FALSE,
  stars_only = FALSE,
  name = "BF",
  protect_ratio = FALSE,
  na_reference = NA,
  exact = FALSE
)
```

Arguments

bf	Bayes Factor.
stars	Add significance stars (e.g., p < .001***).
stars_only	Return only significance stars.
name	Name prefixing the text. Can be NULL.
protect_ratio	Should values smaller than 1 be represented as ratios?
na_reference	How to format missing values (NA).
exact	Should very large or very small values be reported with a scientific format (e.g., 4.24e5), or as truncated values (as "> 1000" and "< 1/1000").

Value

A formatted string.

Examples

```
format_bf(bfs <- c(0.000045, 0.033, NA, 1557, 3.54))
format_bf(bfs, exact = TRUE, name = NULL)
format_bf(bfs, stars = TRUE)
format_bf(bfs, protect_ratio = TRUE)
format_bf(bfs, protect_ratio = TRUE, exact = TRUE)
format_bf(bfs, na_reference = 1)
```

`format_capitalize` *Capitalizes the first letter in a string*

Description

This function converts the first letter in a string into upper case.

Usage

```
format_capitalize(x, verbose = TRUE)
```

Arguments

`x`	A character vector or a factor. The latter is coerced to character. All other objects are returned unchanged.
`verbose`	Toggle warnings.

Value

`x`, with first letter capitalized.

Examples

```
format_capitalize("hello")
format_capitalize(c("hello", "world"))
unique(format_capitalize(iris$Species))
```

`format_ci` *Confidence/Credible Interval (CI) Formatting*

Description

Confidence/Credible Interval (CI) Formatting

Usage

```
format_ci(
  CI_low,
  CI_high,
  ci = 0.95,
  digits = 2,
  brackets = TRUE,
  width = NULL,
  width_low = width,
  width_high = width,
  missing = "",
  zap_small = FALSE
)
```

Arguments

- `CI_low` — Lower CI bound.
- `CI_high` — Upper CI bound.
- `ci` — CI level in percentage.
- `digits` — Number of digits for rounding or significant figures. May also be `"signif"` to return significant figures or `"scientific"` to return scientific notation. Control the number of digits by adding the value as suffix, e.g. `digits = "scientific4"` to have scientific notation with 4 decimal places, or `digits = "signif5"` for 5 significant figures (see also `signif()`).
- `brackets` — Either a logical, and if `TRUE` (default), values are encompassed in square brackets. If `FALSE` or `NULL`, no brackets are used. Else, a character vector of length two, indicating the opening and closing brackets.
- `width` — Minimum width of the returned string. If not `NULL` and `width` is larger than the string's length, leading whitespaces are added to the string. If `width="auto"`, width will be set to the length of the longest string.
- `width_low, width_high` — Like `width`, but only applies to the lower or higher confidence interval value. This can be used when the values for the lower and upper CI are of very different length.
- `missing` — Value by which `NA` values are replaced. By default, an empty string (i.e. `""`) is returned for `NA`.
- `zap_small` — Logical, if `TRUE`, small values are rounded after `digits` decimal places. If `FALSE`, values with more decimal places than `digits` are printed in scientific notation.

Value

A formatted string.

Examples

```
format_ci(1.20, 3.57, ci = 0.90)
format_ci(1.20, 3.57, ci = NULL)
format_ci(1.20, 3.57, ci = NULL, brackets = FALSE)
format_ci(1.20, 3.57, ci = NULL, brackets = c("(", ")"))
format_ci(c(1.205645, 23.4), c(3.57, -1.35), ci = 0.90)
format_ci(c(1.20, NA, NA), c(3.57, -1.35, NA), ci = 0.90)

# automatic alignment of width, useful for printing multiple CIs in columns
x <- format_ci(c(1.205, 23.4, 100.43), c(3.57, -13.35, 9.4))
cat(x, sep = "\n")

x <- format_ci(c(1.205, 23.4, 100.43), c(3.57, -13.35, 9.4), width = "auto")
cat(x, sep = "\n")
```

format_message — *Format messages and warnings*

Description

Inserts line breaks into a longer message or warning string. Line length is adjusted to maximum length of the console, if the width can be accessed. By default, new lines are indented by two whitespace.

Usage

```
format_message(string, ..., line_length = 0.9 * options()$width)
```

Arguments

`string`	A string.
`...`	Further strings that will be concatenated as indented new lines.
`line_length`	Numeric, the maximum length of a line.

Value

A formatted string.

Examples

```
msg <- format_message("Much too long string for just one line, I guess!",
  line_length = 15
)
message(msg)

msg <- format_message("Much too long string for just one line, I guess!",
  "First new line",
  "Second new line",
  "(both indented)",
  line_length = 30
)
message(msg)
```

format_number *Convert number to words*

Description

Convert number to words

Usage

```
format_number(x, textual = TRUE, ...)
```

Arguments

`x`	Number.
`textual`	Return words. If `FALSE`, will run `format_value()`.
`...`	Arguments to be passed to `format_value()` if `textual` is `FALSE`.

Value

A formatted string.

Note

The code has been adapted from here https://github.com/ateucher/useful_code/blob/master/R/numbers2words.r

Examples

```
format_number(2)
format_number(45)
format_number(324.68765)
```

format_p *p-values formatting*

Description

Format p-values.

Usage

```
format_p(
  p,
  stars = FALSE,
  stars_only = FALSE,
  whitespace = TRUE,
  name = "p",
  missing = "",
  decimal_separator = NULL,
  digits = 3,
  ...
)
```

Arguments

`p`	value or vector of p-values.
`stars`	Add significance stars (e.g., p < .001***).
`stars_only`	Return only significance stars.
`whitespace`	Logical, if `TRUE` (default), preserves whitespaces. Else, all whitespace characters are removed from the returned string.
`name`	Name prefixing the text. Can be `NULL`.
`missing`	Value by which `NA` values are replaced. By default, an empty string (i.e. `""`) is returned for `NA`.
`decimal_separator`	Character, if not `NULL`, will be used as decimal separator.
`digits`	Number of significant digits. May also be `"scientific"` to return exact p-values in scientific notation, or `"apa"` to use an APA 7th edition-style for p-values (equivalent to `digits = 3`). If `"scientific"`, control the number of digits by adding the value as a suffix, e.g.m `digits = "scientific4"` to have scientific notation with 4 decimal places.
`...`	Arguments from other methods.

Value

A formatted string.

Examples

```
format_p(c(.02, .065, 0, .23))
format_p(c(.02, .065, 0, .23), name = NULL)
format_p(c(.02, .065, 0, .23), stars_only = TRUE)

model <- lm(mpg ~ wt + cyl, data = mtcars)
p <- coef(summary(model))[, 4]
format_p(p, digits = "apa")
format_p(p, digits = "scientific")
format_p(p, digits = "scientific2")
```

`format_pd` *Probability of direction (pd) formatting*

Description

Probability of direction (pd) formatting

Usage

```
format_pd(pd, stars = FALSE, stars_only = FALSE, name = "pd")
```

Arguments

`pd`	Probability of direction (pd).
`stars`	Add significance stars (e.g., p < .001***).
`stars_only`	Return only significance stars.
`name`	Name prefixing the text. Can be `NULL`.

Value

A formatted string.

Examples

```
format_pd(0.12)
format_pd(c(0.12, 1, 0.9999, 0.98, 0.995, 0.96), name = NULL)
format_pd(c(0.12, 1, 0.9999, 0.98, 0.995, 0.96), stars = TRUE)
```

`format_rope` *Percentage in ROPE formatting*

Description

Percentage in ROPE formatting

Usage

```
format_rope(rope_percentage, name = "in ROPE", digits = 2)
```

Arguments

rope_percentage
: Value or vector of percentages in ROPE.

name
: Name prefixing the text. Can be `NULL`.

digits
: Number of significant digits. May also be `"scientific"` to return exact p-values in scientific notation, or `"apa"` to use an APA 7th edition-style for p-values (equivalent to `digits = 3`). If `"scientific"`, control the number of digits by adding the value as a suffix, e.g.m `digits = "scientific4"` to have scientific notation with 4 decimal places.

Value

A formatted string.

Examples

```
format_rope(c(0.02, 0.12, 0.357, 0))
format_rope(c(0.02, 0.12, 0.357, 0), name = NULL)
```

---

format_string *String Values Formatting*

Description

String Values Formatting

Usage

```
format_string(x, ...)

## S3 method for class 'character'
format_string(x, length = NULL, abbreviate = "...", ...)
```

Arguments

x
: String value.

...
: Arguments passed to or from other methods.

length
: Numeric, maximum length of the returned string. If not `NULL`, will shorten the string to a maximum `length`, however, it will not truncate inside words. I.e. if the string length happens to be inside a word, this word is removed from the returned string, so the returned string has a *maximum* length of `length`, but might be shorter.

abbreviate
: String that will be used as suffix, if `x` was shortened.

Value

A formatted string.

Examples

```
s <- "This can be considered as very long string!"
# string is shorter than max.length, so returned as is
format_string(s, 60)

# string is shortened to as many words that result in
# a string of maximum 20 chars
format_string(s, 20)
```

`format_table` *Parameter table formatting*

Description

This functions takes a data frame with model parameters as input and formats certain columns into a more readable layout (like collapsing separate columns for lower and upper confidence interval values). Furthermore, column names are formatted as well. Note that `format_table()` converts all columns into character vectors!

Usage

```
format_table(
  x,
  pretty_names = TRUE,
  stars = FALSE,
  digits = 2,
  ci_width = "auto",
  ci_brackets = TRUE,
  ci_digits = 2,
  p_digits = 3,
  rope_digits = 2,
  zap_small = FALSE,
  preserve_attributes = FALSE,
  exact = TRUE,
  verbose = TRUE,
  ...
)
```

Arguments

`x`	A data frame of model's parameters, as returned by various functions of the **easystats**-packages. May also be a result from `broom::tidy()`.
`pretty_names`	Return "pretty" (i.e. more human readable) parameter names.
`stars`	Add significance stars (e.g., p < .001***).

digits, ci_digits, p_digits, rope_digits
: Number of digits for rounding or significant figures. May also be "signif" to return significant figures or "scientific" to return scientific notation. Control the number of digits by adding the value as suffix, e.g. digits = "scientific4" to have scientific notation with 4 decimal places, or digits = "signif5" for 5 significant figures (see also signif()).

ci_width
: Minimum width of the returned string for confidence intervals. If not NULL and width is larger than the string's length, leading whitespaces are added to the string. If width="auto", width will be set to the length of the longest string.

ci_brackets
: Logical, if TRUE (default), CI-values are encompassed in square brackets (else in parentheses).

zap_small
: Logical, if TRUE, small values are rounded after digits decimal places. If FALSE, values with more decimal places than digits are printed in scientific notation.

preserve_attributes
: Logical, if TRUE, preserves all attributes from the input data frame.

exact
: Formatting for Bayes factor columns, in case the provided data frame contains such a column (i.e. columns named "BF" or "log_BF"). For exact = TRUE, very large or very small values are then either reported with a scientific format (e.g., 4.24e5), else as truncated values (as "> 1000" and "< 1/1000").

verbose
: Toggle messages and warnings.

...
: Arguments passed to or from other methods.

Value

A data frame. Note that format_table() converts all columns into character vectors!

See Also

Vignettes Formatting, printing and exporting tables and Formatting model parameters.

Examples

```
format_table(head(iris), digits = 1)

if (require("parameters")) {
  x <- model_parameters(lm(Sepal.Length ~ Species * Sepal.Width, data = iris))
  as.data.frame(format_table(x))
  as.data.frame(format_table(x, p_digits = "scientific"))
}

if (require("rstanarm", warn.conflicts = FALSE) &&
  require("parameters", , warn.conflicts = FALSE)) {
  model <- stan_glm(Sepal.Length ~ Species, data = iris, refresh = 0, seed = 123)
  x <- model_parameters(model, ci = c(0.69, 0.89, 0.95))
  as.data.frame(format_table(x))
}
```

`format_value` *Numeric Values Formatting*

Description

`format_value()` converts numeric values into formatted string values2, where formatting can be something like rounding digits, scientific notation etc. `format_percent()` is a short-cut for `format_value(as_percent = TRUE)`.

Usage

```
format_value(x, ...)

## S3 method for class 'data.frame'
format_value(
  x,
  digits = 2,
  protect_integers = FALSE,
  missing = "",
  width = NULL,
  as_percent = FALSE,
  zap_small = FALSE,
  ...
)

## S3 method for class 'numeric'
format_value(
  x,
  digits = 2,
  protect_integers = FALSE,
  missing = "",
  width = NULL,
  as_percent = FALSE,
  zap_small = FALSE,
  ...
)

format_percent(x, ...)
```

Arguments

`x`	Numeric value.
`...`	Arguments passed to or from other methods.
`digits`	Number of digits for rounding or significant figures. May also be `"signif"` to return significant figures or `"scientific"` to return scientific notation. Control the number of digits by adding the value as suffix, e.g. `digits = "scientific4"`

to have scientific notation with 4 decimal places, or `digits = "signif5"` for 5 significant figures (see also `signif()`).

`protect_integers`
: Should integers be kept as integers (i.e., without decimals)?

`missing`
: Value by which `NA` values are replaced. By default, an empty string (i.e. `""`) is returned for `NA`.

`width`
: Minimum width of the returned string. If not `NULL` and `width` is larger than the string's length, leading whitespaces are added to the string.

`as_percent`
: Logical, if `TRUE`, value is formatted as percentage value.

`zap_small`
: Logical, if `TRUE`, small values are rounded after `digits` decimal places. If `FALSE`, values with more decimal places than `digits` are printed in scientific notation.

Value

A formatted string.

Examples

```
format_value(1.20)
format_value(1.2)
format_value(1.2012313)
format_value(c(0.0045, 234, -23))
format_value(c(0.0045, .12, .34))
format_value(c(0.0045, .12, .34), as_percent = TRUE)
format_value(c(0.0045, .12, .34), digits = "scientific")
format_value(c(0.0045, .12, .34), digits = "scientific2")

# default
format_value(c(0.0045, .123, .345))
# significant figures
format_value(c(0.0045, .123, .345), digits = "signif")

format_value(as.factor(c("A", "B", "A")))
format_value(iris$Species)

format_value(3)
format_value(3, protect_integers = TRUE)

format_value(head(iris))
```

`get_auxiliary` *Get auxiliary parameters from models*

Description

Returns the requested auxiliary parameters from models, like dispersion, sigma, or beta...

Usage

```
get_auxiliary(
  x,
  type = "sigma",
  summary = TRUE,
  centrality = "mean",
  verbose = TRUE,
  ...
)
```

Arguments

`x`	A model.
`type`	The name of the auxiliary parameter that should be retrieved. `"sigma"` is available for most models, `"dispersion"` for models of class `glm`, `glmerMod` or `glmmTMB` as well as `brmsfit`. `"beta"` and other parameters are currently only returned for `brmsfit` models. See 'Details'.
`summary`	Logical, indicates whether the full posterior samples (`summary = FALSE`)) or the summarized centrality indices of the posterior samples (`summary = TRUE`)) should be returned as estimates.
`centrality`	Only for models with posterior samples, and when `summary = TRUE`. In this case, `centrality = "mean"` would calculate means of posterior samples for each parameter, while `centrality = "median"` would use the more robust median value as measure of central tendency.
`verbose`	Toggle warnings.
`...`	Currently not used.

Details

Currently, only sigma and the dispersion parameter are returned, and only for a limited set of models.

Sigma Parameter: See `get_sigma()`.

Dispersion Parameter: There are many different definitions of "dispersion", depending on the context. `get_auxiliary()` returns the dispersion parameters that usually can be considered as variance-to-mean ratio for generalized (linear) mixed models. Exceptions are models of class `glmmTMB`, where the dispersion equals σ^2. In detail, the computation of the dispersion parameter for generalized linear models is the ratio of the sum of the squared working-residuals and the residual degrees of freedom. For mixed models of class `glmer`, the dispersion parameter is also called ϕ and is the ratio of the sum of the squared Pearson-residuals and the residual degrees of freedom. For models of class `glmmTMB`, dispersion is σ^2.

brms models: For models of class `brmsfit`, there are different options for the `type` argument. See a list of supported auxiliary parameters here: `find_parameters.BGGM()`.

Value

The requested auxiliary parameter, or `NULL` if this information could not be accessed.

Examples

```
# from ?glm
clotting <- data.frame(
  u = c(5, 10, 15, 20, 30, 40, 60, 80, 100),
  lot1 = c(118, 58, 42, 35, 27, 25, 21, 19, 18),
  lot2 = c(69, 35, 26, 21, 18, 16, 13, 12, 12)
)
model <- glm(lot1 ~ log(u), data = clotting, family = Gamma())
get_auxiliary(model, type = "dispersion") # same as summary(model)$dispersion
```

`get_call` *Get the model's function call*

Description

Returns the model's function call when available.

Usage

```
get_call(x)
```

Arguments

`x`	A fitted mixed model.

Value

A function call.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_call(m)

if (require("lme4")) {
  m <- lmer(Sepal.Length ~ Sepal.Width + (1 | Species), data = iris)
  get_call(m)
}
```

get_data *Get the data that was used to fit the model*

Description

This functions tries to get the data that was used to fit the model and returns it as data frame.

Usage

```
get_data(x, ...)

## Default S3 method:
get_data(x, verbose = TRUE, ...)

## S3 method for class 'glmmTMB'
get_data(x, effects = "all", component = "all", verbose = TRUE, ...)

## S3 method for class 'afex_aov'
get_data(x, shape = c("long", "wide"), ...)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.
`verbose`	Toggle messages and warnings.
`effects`	Should model data for fixed effects (`"fixed"`), random effects (`"random"`) or both (`"all"`) be returned? Only applies to mixed or gee models.
`component`	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
`shape`	Return long or wide data? Only applicable in repeated measures designs.

Value

The data that was used to fit the model.

Model components

Possible values for the `component` argument depend on the model class. Following are valid options:

- `"all"`: returns all model components, applies to all models, but will only have an effect for models with more than just the conditional model component.

- "conditional": only returns the conditional component, i.e. "fixed effects" terms from the model. Will only have an effect for models with more than just the conditional model component.
- "smooth_terms": returns smooth terms, only applies to GAMs (or similar models that may contain smooth terms).
- "zero_inflated" (or "zi"): returns the zero-inflation component.
- "dispersion": returns the dispersion model component. This is common for models with zero-inflation or that can model the dispersion parameter.
- "instruments": for instrumental-variable or some fixed effects regression, returns the instruments.
- "location": returns location parameters such as conditional, zero_inflated, smooth_terms, or instruments (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters).
- "distributional" (or "auxiliary"): components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.

Note

Unlike model.frame(), which may contain transformed variables (e.g. if poly() or scale() was used inside the formula to specify the model), get_data() aims at returning the "original", untransformed data (if possible). Consequently, column names are changed accordingly, i.e. "log(x)" will become "x" etc. for all data columns with transformed values.

Examples

```
if (require("lme4")) {
  data(cbpp, package = "lme4")
  cbpp$trials <- cbpp$size - cbpp$incidence
  m <- glm(cbind(incidence, trials) ~ period, data = cbpp, family = binomial)
  head(get_data(m))
}
```

---

get_datagrid *Create a reference grid*

Description

Create a reference matrix, useful for visualisation, with evenly spread and combined values. Usually used to make generate predictions using get_predicted(). See this vignette for a tutorial on how to create a visualisation matrix using this function.

Usage

```
get_datagrid(x, ...)

## S3 method for class 'data.frame'
get_datagrid(
  x,
  at = "all",
  factors = "reference",
  numerics = "mean",
  preserve_range = FALSE,
  reference = x,
  length = 10,
  range = "range",
  ...
)

## S3 method for class 'numeric'
get_datagrid(x, length = 10, range = "range", ...)

## S3 method for class 'factor'
get_datagrid(x, ...)

## Default S3 method:
get_datagrid(
  x,
  at = "all",
  factors = "reference",
  numerics = "mean",
  preserve_range = TRUE,
  reference = x,
  include_smooth = TRUE,
  include_random = FALSE,
  include_response = FALSE,
  data = NULL,
  ...
)
```

Arguments

`x`	An object from which to construct the reference grid.
`...`	Arguments passed to or from other methods (for instance, `length` or `range` to control the spread of numeric variables.).
`at`	Can be `"all"`, a character vector or list of named elements, indicating the predictors of interest (focal predictors). Can also contain assignments (as named list, e.g. `at = list(c(Sepal.Length = c(2, 4), Species = "setosa"))`, or as string, e.g. `at = "Sepal.Length = 2"` or `at = c("Sepal.Length = 2", "Species = 'setosa'")` - note the usage of single and double quotes to assign strings within strings). The remaining variables will be fixed.

factors	Type of summary for factors. Can be "reference" (set at the reference level), "mode" (set at the most common level) or "all" to keep all levels.
numerics	Type of summary for numeric values. Can be "all" (will duplicate the grid for all unique values), any function ("mean", "median", ...) or a value (e.g., numerics = 0).
preserve_range	In the case of combinations between numeric variables and factors, setting preserve_range = TRUE will drop the observations where the value of the numeric variable is originally not present in the range of its factor level. This leads to an unbalanced grid. Also, if you want the minimum and the maximum to closely match the actual ranges, you should increase the length argument.
reference	The reference vector from which to compute the mean and SD. Used when standardizing or unstandardizing the grid using effectsize::standardize.
length	Length of numeric target variables selected in "at". This arguments controls the number of (equally spread) values that will be taken to represent the continuous variables. A longer length will increase precision, but can also substantially increase the size of the datagrid (especially in case of interactions). If NA, will return all the unique values. In case of multiple continuous target variables, length can also be a vector of different values (see examples).
range	If "range" (default), will use the minimum and maximum of the original data vector as end-points (min and max). If an interval type is specified, such as "iqr", "ci", "hdi" or "eti", it will spread the values within that range (the default CI width is 95% but this can be changed by adding for instance ci = 0.90. See IQR() and bayestestR::ci(). This can be useful to have more robust change and skipping extreme values. If "sd" or "mad", it will spread by this dispersion index around the mean or the median, respectively. If the length argument is an even number (e.g., 4), it will have one more step on the positive side (i.e., -1, 0, +1, +2). The result is a named vector. See examples.
include_smooth	If x is a model object, decide whether smooth terms should be included in the data grid or not.
include_random	If x is a mixed model object, decide whether random effect terms should be included in the data grid or not. If include_random is FALSE, but x is a mixed model with random effects, these will still be included in the returned grid, but set to their "population level" value (e.g., NA for *glmmTMB* or 0 for *merMod*). This ensures that common predict() methods work properly, as these usually need data with all variables in the model included.
include_response	If x is a model object, decide whether the response variable should be included in the data grid or not.
data	Optional, the data frame that was used to fit the model. Usually, the data is retrieved via get_data().

Value

Reference grid data frame.

See Also

get_predicted()

Examples

```
# Datagrids of variables and dataframes =====================================
if (require("bayestestR", quietly = TRUE) & require("datawizard", quietly = TRUE)) {
  # Single variable is of interest; all others are "fixed" ------------------
  # Factors
  get_datagrid(iris, at = "Species") # Returns all the levels
  get_datagrid(iris, at = "Species = c('setosa', 'versicolor')") # Specify an expression

  # Numeric variables
  get_datagrid(iris, at = "Sepal.Length") # default spread length = 10
  get_datagrid(iris, at = "Sepal.Length", length = 3) # change length
  get_datagrid(iris[2:150, ],
    at = "Sepal.Length",
    factors = "mode", numerics = "median"
  ) # change non-targets fixing
  get_datagrid(iris, at = "Sepal.Length", range = "ci", ci = 0.90) # change min/max of target
  get_datagrid(iris, at = "Sepal.Length = [0, 1]") # Manually change min/max


  # Standardization and unstandardization
  data <- get_datagrid(iris, at = "Sepal.Length", range = "sd", length = 3)
  data$Sepal.Length # It is a named vector (extract names with `names(out$Sepal.Length)`)
  # TODO: uncomment when datawizard > 0.3.1 is out
  # datawizard::standardize(data, select = "Sepal.Length")
  # data <- get_datagrid(iris, at = "Sepal.Length = c(-2, 0, 2)") # Manually specify values
  # data
  # datawizard::unstandardize(data, select = "Sepal.Length")


  # Multiple variables are of interest, creating a combination --------------
  get_datagrid(iris, at = c("Sepal.Length", "Species"), length = 3)
  get_datagrid(iris, at = c("Sepal.Length", "Petal.Length"), length = c(3, 2))
  get_datagrid(iris, at = c(1, 3), length = 3)
  get_datagrid(iris, at = c("Sepal.Length", "Species"), preserve_range = TRUE)
  get_datagrid(iris, at = c("Sepal.Length", "Species"), numerics = 0)
  get_datagrid(iris, at = c("Sepal.Length = 3", "Species"))
  get_datagrid(iris, at = c("Sepal.Length = c(3, 1)", "Species = 'setosa'"))

  # With list-style at-argument
  get_datagrid(iris, at = list(Sepal.Length = c(1, 3), Species = "setosa"))
}

# With models ===============================================================
# Fit a linear regression
model <- lm(Sepal.Length ~ Sepal.Width * Petal.Length, data = iris)
# Get datagrid of predictors
data <- get_datagrid(model, length = c(20, 3), range = c("range", "sd"))
# Add predictions
data$Sepal.Length <- get_predicted(model, data = data)
# Visualize relationships (each color is at -1 SD, Mean, and + 1 SD of Petal.Length)
plot(data$Sepal.Width, data$Sepal.Length,
  col = data$Petal.Length,
```

```
  main = "Relationship at -1 SD, Mean, and + 1 SD of Petal.Length"
)
```

---

get_deviance *Model Deviance*

Description

Returns model deviance (see `stats::deviance()`).

Usage

```
get_deviance(x, ...)

## Default S3 method:
get_deviance(x, verbose = TRUE, ...)
```

Arguments

x	A model.
...	Not used.
verbose	Toggle warnings and messages.

Details

For GLMMs of class `glmerMod`, `glmmTMB` or `MixMod`, the *absolute unconditional* deviance is returned (see 'Details' in `?lme4::merMod-class`), i.e. minus twice the log-likelihood. To get the *relative conditional* deviance (relative to a saturated model, conditioned on the conditional modes of random effects), use `deviance()`. The value returned `get_deviance()` usually equals the deviance-value from the `summary()`.

Value

The model deviance.

Examples

```
data(mtcars)
x <- lm(mpg ~ cyl, data = mtcars)
get_deviance(x)
```

get_df *Extract degrees of freedom*

Description

Estimate or extract residual or model-based degrees of freedom from regression models.

Usage

```
get_df(x, ...)

## Default S3 method:
get_df(x, type = "residual", verbose = TRUE, ...)
```

Arguments

`x`	A statistical model.
`...`	Currently not used.
`type`	Can be `"residual"`, `"model"` or `"analytical"`. `"residual"` tries to extract residual degrees of freedoms. If residual degrees of freedom could not be extracted, returns analytical degrees of freedom, i.e. `n-k` (number of observations minus number of parameters). `"model"` returns model-based degrees of freedom, i.e. the number of (estimated) parameters.
`verbose`	Toggle warnings.

Examples

```
model <- lm(Sepal.Length ~ Petal.Length * Species, data = iris)
get_df(model) # same as df.residual(model)
get_df(model, type = "model") # same as attr(logLik(model), "df")
```

get_family *A robust alternative to stats::family*

Description

A robust and resilient alternative to `stats::family`. To avoid issues with models like `gamm4`.

Usage

```
get_family(x, ...)
```

Arguments

`x`	A statistical model.
`...`	Further arguments passed to methods.

Examples

```
data(mtcars)
x <- glm(vs ~ wt, data = mtcars, family = "binomial")
get_family(x)

if (require("mgcv")) {
  x <- mgcv::gamm(
    vs ~ am + s(wt),
    random = list(cyl = ~1),
    data = mtcars,
    family = "binomial"
  )
  get_family(x)
}
```

`get_intercept` *Get the value at the intercept*

Description

Returns the value at the intercept (i.e., the intercept parameter), and `NA` if there isn't one.

Usage

```
get_intercept(x, ...)
```

Arguments

`x`	A model.
`...`	Not used.

Value

The value of the intercept.

Examples

```
get_intercept(lm(Sepal.Length ~ Petal.Width, data = iris))
get_intercept(lm(Sepal.Length ~ 0 + Petal.Width, data = iris))

if (require("lme4")) {
  get_intercept(lme4::lmer(Sepal.Length ~ Sepal.Width + (1 | Species), data = iris))
}
if (require("gamm4")) {
  get_intercept(gamm4::gamm4(Sepal.Length ~ s(Petal.Width), data = iris))
}
```

get_loglikelihood *Log-Likelihood*

Description

A robust function to compute the log-likelihood of a model, as well as individual log-likelihoods (for each observation) whenever possible. Can be used as a replacement for `stats::logLik()` out of the box, as the returned object is of the same class (and it gives the same results by default).

Usage

```
get_loglikelihood(x, ...)

loglikelihood(x, ...)

## S3 method for class 'lm'
get_loglikelihood(
  x,
  estimator = "ML",
  REML = FALSE,
  check_response = FALSE,
  verbose = TRUE,
  ...
)
```

Arguments

`x`	A model.
`...`	Passed down to `logLik()`, if possible.
`estimator`	Corresponds to the different estimators for the standard deviation of the errors. If `estimator="ML"` (default), the scaling is done by n (the biased ML estimator), which is then equivalent to using `stats::logLik()`. If `estimator="OLS"`, it returns the unbiased OLS estimator. `estimator="REML"` will give same results as `logLik(..., REML=TRUE)`.
`REML`	Only for linear models. This argument is present for compatibility with `stats::logLik()`. Setting it to `TRUE` will overwrite the `estimator` argument and is thus equivalent to setting `estimator="REML"`. It will give the same results as `stats::logLik(..., REML=TRUE)`. Note that individual log-likelihoods are not available under REML.
`check_response`	Logical, if `TRUE`, checks if the response variable is transformed (like `log()` or `sqrt()`), and if so, returns a corrected log-likelihood. To get back to the original scale, the likelihood of the model is multiplied by the Jacobian/derivative of the transformation.
`verbose`	Toggle warnings and messages.

Value

An object of class "logLik", also containing the log-likelihoods for each observation as a per_observation attribute (attributes(get_loglikelihood(x))$per_observation) when possible. The code was partly inspired from the **nonnest2** package.

Examples

```
x <- lm(Sepal.Length ~ Petal.Width + Species, data = iris)

get_loglikelihood(x, estimator = "ML") # Equivalent to stats::logLik(x)
get_loglikelihood(x, estimator = "REML") # Equivalent to stats::logLik(x, REML=TRUE)
get_loglikelihood(x, estimator = "OLS")
```

---

get_modelmatrix *Model Matrix*

Description

Creates a design matrix from the description. Any character variables are coerced to factors.

Usage

```
get_modelmatrix(x, ...)
```

Arguments

x	An object.
...	Passed down to other methods (mainly model.matrix()).

Examples

```
data(mtcars)

model <- lm(am ~ vs, data = mtcars)
get_modelmatrix(model)
```

get_parameters *Get model parameters*

Description

Returns the coefficients (or posterior samples for Bayesian models) from a model. See the documentation for your object's class:

- Bayesian models (**rstanarm**, **brms**, **MCMCglmm**, ...)
- Estimated marginal means (**emmeans**)
- Generalized additive models (**mgcv**, **VGAM**, ...)
- Marginal effects models (**mfx**)
- Mixed models (**lme4**, **glmmTMB**, **GLMMadaptive**, ...)
- Zero-inflated and hurdle models (**pscl**, ...)
- Models with special components (**betareg**, **MuMIn**, ...)
- Hypothesis tests (`htest`)

Usage

```
get_parameters(x, ...)

## Default S3 method:
get_parameters(x, verbose = TRUE, ...)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.
`verbose`	Toggle messages and warnings.

Details

In most cases when models either return different "effects" (fixed, random) or "components" (conditional, zero-inflated, ...), the arguments `effects` and `component` can be used.

`get_parameters()` is comparable to `coef()`, however, the coefficients are returned as data frame (with columns for names and point estimates of coefficients). For Bayesian models, the posterior samples of parameters are returned.

Value

- for non-Bayesian models, a data frame with two columns: the parameter names and the related point estimates.
- for Anova (`aov()`) with error term, a list of parameters for the conditional and the random effects parameters

Model components

Possible values for the component argument depend on the model class. Following are valid options:

- "all": returns all model components, applies to all models, but will only have an effect for models with more than just the conditional model component.
- "conditional": only returns the conditional component, i.e. "fixed effects" terms from the model. Will only have an effect for models with more than just the conditional model component.
- "smooth_terms": returns smooth terms, only applies to GAMs (or similar models that may contain smooth terms).
- "zero_inflated" (or "zi"): returns the zero-inflation component.
- "dispersion": returns the dispersion model component. This is common for models with zero-inflation or that can model the dispersion parameter.
- "instruments": for instrumental-variable or some fixed effects regression, returns the instruments.
- "location": returns location parameters such as conditional, zero_inflated, smooth_terms, or instruments (everything that are fixed or random effects - depending on the effects argument - but no auxiliary parameters).
- "distributional" (or "auxiliary"): components like sigma, dispersion, beta or precision (and other auxiliary parameters) are returned.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

---

get_parameters.betamfx

Get model parameters from marginal effects models

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'betamfx'
get_parameters(
  x,
  component = c("all", "conditional", "precision", "marginal"),
  ...
)

## S3 method for class 'logitmfx'
get_parameters(x, component = c("all", "conditional", "marginal"), ...)
```

Arguments

- `x` A fitted model.
- `component` Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
- `...` Currently not used.

Value

A data frame with three columns: the parameter names, the related point estimates and the component.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

---

get_parameters.betareg

Get model parameters from models with special components

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'betareg'
get_parameters(
  x,
  component = c("all", "conditional", "precision", "location", "distributional",
    "auxiliary"),
  ...
)

## S3 method for class 'DirichletRegModel'
get_parameters(
  x,
  component = c("all", "conditional", "precision", "location", "distributional",
    "auxiliary"),
  ...
)
```

```
## S3 method for class 'averaging'
get_parameters(x, component = c("conditional", "full"), ...)

## S3 method for class 'glmx'
get_parameters(
  x,
 component = c("all", "conditional", "extra", "location", "distributional", "auxiliary"),
  ...
)

## S3 method for class 'clm2'
get_parameters(x, component = c("all", "conditional", "scale"), ...)

## S3 method for class 'mvord'
get_parameters(
  x,
  component = c("all", "conditional", "thresholds", "correlation"),
  ...
)

## S3 method for class 'mjoint'
get_parameters(x, component = c("all", "conditional", "survival"), ...)
```

Arguments

x	A fitted model.
component	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
...	Currently not used.

Value

A data frame with three columns: the parameter names, the related point estimates and the component.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

get_parameters.BGGM *Get model parameters from Bayesian models*

Description

Returns the coefficients (or posterior samples for Bayesian models) from a model.

Usage

```
## S3 method for class 'BGGM'
get_parameters(
  x,
  component = c("correlation", "conditional", "intercept", "all"),
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'MCMCglmm'
get_parameters(
  x,
  effects = c("fixed", "random", "all"),
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'BFBayesFactor'
get_parameters(
  x,
  effects = c("all", "fixed", "random"),
  component = c("all", "extra"),
  iterations = 4000,
  progress = FALSE,
  verbose = TRUE,
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'stanmvreg'
get_parameters(
  x,
  effects = c("fixed", "random", "all"),
  parameters = NULL,
  summary = FALSE,
  centrality = "mean",
```

```
  ...
)

## S3 method for class 'brmsfit'
get_parameters(
  x,
  effects = "fixed",
  component = "all",
  parameters = NULL,
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'stanreg'
get_parameters(
  x,
  effects = c("fixed", "random", "all"),
  component = c("location", "all", "conditional", "smooth_terms", "sigma",
    "distributional", "auxiliary"),
  parameters = NULL,
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'bayesx'
get_parameters(
  x,
  component = c("conditional", "smooth_terms", "all"),
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'bamlss'
get_parameters(
  x,
  component = c("all", "conditional", "smooth_terms", "location", "distributional",
    "auxiliary"),
  parameters = NULL,
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'sim.merMod'
get_parameters(
```

```
  x,
  effects = c("fixed", "random", "all"),
  parameters = NULL,
  summary = FALSE,
  centrality = "mean",
  ...
)

## S3 method for class 'sim'
get_parameters(x, parameters = NULL, summary = FALSE, centrality = "mean", ...)
```

Arguments

`x`	A fitted model.
`component`	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model, the dispersion term, the instrumental variables or marginal effects be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variables (so called fixed-effects regressions), or models with marginal effects from **mfx**. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional`, `zero_inflated`, `smooth_terms`, or `instruments` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma`, `dispersion`, `beta` or `precision` (and other auxiliary parameters) are returned.
`summary`	Logical, indicates whether the full posterior samples (`summary = FALSE`)) or the summarized centrality indices of the posterior samples (`summary = TRUE`)) should be returned as estimates.
`centrality`	Only for models with posterior samples, and when `summary = TRUE`. In this case, `centrality = "mean"` would calculate means of posterior samples for each parameter, while `centrality = "median"` would use the more robust median value as measure of central tendency.
`...`	Currently not used.
`effects`	Should parameters for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`iterations`	Number of posterior draws.
`progress`	Display progress.
`verbose`	Toggle messages and warnings.
`parameters`	Regular expression pattern that describes the parameters that should be returned.

Details

In most cases when models either return different "effects" (fixed, random) or "components" (conditional, zero-inflated, ...), the arguments `effects` and `component` can be used.

Value

The posterior samples from the requested parameters as data frame. If `summary = TRUE`, returns a data frame with two columns: the parameter names and the related point estimates (based on `centrality`).

BFBayesFactor Models

Note that for `BFBayesFactor` models (from the **BayesFactor** package), posteriors are only extracted from the first numerator model (i.e., `model[1]`). If you want to apply some function `foo()` to another model stored in the `BFBayesFactor` object, index it directly, e.g. `foo(model[2])`, `foo(1/model[5])`, etc. See also `bayestestR::weighted_posteriors()`.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

`get_parameters.emmGrid`

Get model parameters from estimated marginal means objects

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'emmGrid'
get_parameters(x, summary = FALSE, merge_parameters = FALSE, ...)

## S3 method for class 'emm_list'
get_parameters(x, summary = FALSE, ...)
```

Arguments

`x`	A fitted model.
`summary`	Logical, indicates whether the full posterior samples (`summary = FALSE`)) or the summarized centrality indices of the posterior samples (`summary = TRUE`)) should be returned as estimates.
`merge_parameters`	Logical, if `TRUE` and `x` has multiple columns for parameter names (like `emmGrid` objects may have), these are merged into a single parameter column, with parameters names and values as values.
`...`	Currently not used.

Value

A data frame with two columns: the parameter names and the related point estimates.

Note

Note that `emmGrid` or `emm_list` objects returned by functions from **emmeans** have a different structure compared to usual regression models. Hence, the `Parameter` column does not always contain names of *variables*, but may rather contain *values*, e.g. for contrasts. See an example for pairwise comparisons below.

Examples

```
data(mtcars)
model <- lm(mpg ~ wt * factor(cyl), data = mtcars)
if (require("emmeans", quietly = TRUE)) {
  emm <- emmeans(model, "cyl")
  get_parameters(emm)

  emm <- emmeans(model, pairwise ~ cyl)
  get_parameters(emm)
}
```

`get_parameters.gamm` *Get model parameters from generalized additive models*

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'gamm'
get_parameters(
  x,
  component = c("all", "conditional", "smooth_terms", "location"),
  ...
)

## S3 method for class 'gam'
get_parameters(
  x,
  component = c("all", "conditional", "smooth_terms", "location"),
  ...
)

## S3 method for class 'rqss'
get_parameters(x, component = c("all", "conditional", "smooth_terms"), ...)
```

Arguments

x	A fitted model.
component	Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
...	Currently not used.

Value

For models with smooth terms or zero-inflation component, a data frame with three columns: the parameter names, the related point estimates and the component.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

---

get_parameters.glmm *Get model parameters from mixed models*

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'glmm'
get_parameters(x, effects = c("all", "fixed", "random"), ...)

## S3 method for class 'coxme'
get_parameters(x, effects = c("fixed", "random"), ...)

## S3 method for class 'merMod'
get_parameters(x, effects = c("fixed", "random"), ...)

## S3 method for class 'glmmTMB'
get_parameters(
  x,
  effects = c("fixed", "random"),
  component = c("all", "conditional", "zi", "zero_inflated", "dispersion"),
  ...
)
```

```
## S3 method for class 'glimML'
get_parameters(x, effects = c("fixed", "random", "all"), ...)
```

Arguments

`x`	A fitted model.
`effects`	Should parameters for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`...`	Currently not used.
`component`	Which type of parameters to return, such as parameters for the conditional model, the zero-inflated part of the model or the dispersion term? Applies to models with zero-inflated and/or dispersion formula. Note that the *conditional* component is also called *count* or *mean* component, depending on the model. There are three convenient shortcuts: `component = "all"` returns all possible parameters. If `component = "location"`, location parameters such as `conditional` or `zero_inflated` are returned (everything that are fixed or random effects - depending on the `effects` argument - but no auxiliary parameters). For `component = "distributional"` (or `"auxiliary"`), components like `sigma` or `dispersion` (and other auxiliary parameters) are returned.

Details

In most cases when models either return different "effects" (fixed, random) or "components" (conditional, zero-inflated, ...), the arguments `effects` and `component` can be used.

Value

If `effects = "fixed"`, a data frame with two columns: the parameter names and the related point estimates. If `effects = "random"`, a list of data frames with the random effects (as returned by `ranef()`), unless the random effects have the same simplified structure as fixed effects (e.g. for models from **MCMCglmm**).

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

`get_parameters.htest` *Get model parameters from htest-objects*

Description

Returns the parameters from a hypothesis test.

Usage

```
## S3 method for class 'htest'
get_parameters(x, ...)
```

Arguments

x	A fitted model.
...	Currently not used.

Value

A data frame with two columns: the parameter names and the related point estimates.

Examples

```
get_parameters(t.test(1:10, y = c(7:20)))
```

---

get_parameters.zeroinfl

Get model parameters from zero-inflated and hurdle models

Description

Returns the coefficients from a model.

Usage

```
## S3 method for class 'zeroinfl'
get_parameters(
  x,
  component = c("all", "conditional", "zi", "zero_inflated"),
  ...
)

## S3 method for class 'zcpglm'
get_parameters(
  x,
  component = c("all", "conditional", "zi", "zero_inflated"),
  ...
)

## S3 method for class 'mhurdle'
get_parameters(
  x,
 component = c("all", "conditional", "zi", "zero_inflated", "infrequent_purchase", "ip",
    "auxiliary"),
  ...
)
```

Arguments

- `x` A fitted model.
- `component` Should all predictor variables, predictor variables for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
- `...` Currently not used.

Value

For models with smooth terms or zero-inflation component, a data frame with three columns: the parameter names, the related point estimates and the component.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_parameters(m)
```

---

get_predicted *Model predictions (robust) and their confidence intervals*

Description

The `get_predicted()` function is a robust, flexible and user-friendly alternative to base R `predict()` function. Additional features and advantages include availability of uncertainty intervals (CI), bootstrapping, a more intuitive API and the support of more models than base R's `predict()` function. However, although the interface are simplified, it is still very important to read the documentation of the arguments. This is because making "predictions" (a lose term for a variety of things) is a non-trivial process, with lots of caveats and complications. Read the 'Details' section for more information.

`get_predicted_ci()` returns the confidence (or prediction) interval (CI) associated with predictions made by a model. This function can be called separately on a vector of predicted values. `get_predicted()` usually returns confidence intervals (included as attribute, and accessible via the `as.data.frame()` method) by default.

Usage

```
get_predicted(x, ...)

## Default S3 method:
get_predicted(
  x,
```

```
  data = NULL,
  predict = "expectation",
  ci = NULL,
  ci_type = "confidence",
  ci_method = NULL,
  dispersion_method = "sd",
  vcov = NULL,
  vcov_args = NULL,
  verbose = TRUE,
  ...
)

## S3 method for class 'lm'
get_predicted(
  x,
  data = NULL,
  predict = "expectation",
  ci = NULL,
  iterations = NULL,
  verbose = TRUE,
  ...
)

## S3 method for class 'stanreg'
get_predicted(
  x,
  data = NULL,
  predict = "expectation",
  iterations = NULL,
  ci = NULL,
  ci_method = NULL,
  include_random = "default",
  include_smooth = TRUE,
  verbose = TRUE,
  ...
)

## S3 method for class 'gam'
get_predicted(
  x,
  data = NULL,
  predict = "expectation",
  ci = NULL,
  include_random = TRUE,
  include_smooth = TRUE,
  iterations = NULL,
  verbose = TRUE,
  ...
```

```
)

## S3 method for class 'lmerMod'
get_predicted(
  x,
  data = NULL,
  predict = "expectation",
  ci = NULL,
  ci_method = NULL,
  include_random = "default",
  iterations = NULL,
  verbose = TRUE,
  ...
)

## S3 method for class 'principal'
get_predicted(x, data = NULL, ...)
```

Arguments

`x` A statistical model (can also be a data.frame, in which case the second argument has to be a model).

`...` Other argument to be passed, for instance to `get_predicted_ci()`.

`data` An optional data frame in which to look for variables with which to predict. If omitted, the data used to fit the model is used. Visualization matrices can be generated using `get_datagrid()`.

`predict` string or `NULL`

- `"link"` returns predictions on the model's link-scale (for logistic models, that means the log-odds scale) with a confidence interval (CI).
- `"expectation"` (default) also returns confidence intervals, but this time the output is on the response scale (for logistic models, that means probabilities).
- `"prediction"` also gives an output on the response scale, but this time associated with a prediction interval (PI), which is larger than a confidence interval (though it mostly make sense for linear models).
- `"classification"` only differs from `"prediction"` for binomial models where it additionally transforms the predictions into the original response's type (for instance, to a factor).
- Other strings are passed directly to the `type` argument of the `predict()` method supplied by the modelling package.
- When `predict = NULL`, alternative arguments such as `type` will be captured by the `...` ellipsis and passed directly to the `predict()` method supplied by the modelling package. Note that this might result in conflicts with multiple matching `type` arguments - thus, the recommendation is to use the `predict` argument for those values.

- Notes: You can see the 4 options for predictions as on a gradient from "close to the model" to "close to the response data": "link", "expectation", "prediction", "classification". The `predict` argument modulates two things: the scale of the output and the type of certainty interval. Read more about in the **Details** section below.

`ci` The interval level. Default is `NULL`, to be fast even for larger models. Set the interval level to an explicit value, e.g. `0.95`, for `95%` CI).

`ci_type` Can be `"prediction"` or `"confidence"`. Prediction intervals show the range that likely contains the value of a new observation (in what range it would fall), whereas confidence intervals reflect the uncertainty around the estimated parameters (and gives the range of the link; for instance of the regression line in a linear regressions). Prediction intervals account for both the uncertainty in the model's parameters, plus the random variation of the individual values. Thus, prediction intervals are always wider than confidence intervals. Moreover, prediction intervals will not necessarily become narrower as the sample size increases (as they do not reflect only the quality of the fit). This applies mostly for "simple" linear models (like `lm`), as for other models (e.g., `glm`), prediction intervals are somewhat useless (for instance, for a binomial model for which the dependent variable is a vector of 1s and 0s, the prediction interval is... `[0, 1]`).

`ci_method` The method for computing p values and confidence intervals. Possible values depend on model type.

- `NULL` uses the default method, which varies based on the model type.
- Most frequentist models: `"normal"` (default).
- Bayesian models: `"quantile"` (default), `"hdi"`, `"eti"`, and `"spi"`.
- Mixed effects **lme4** models: `"normal"` (default), `"satterthwaite"`, and `"kenward"`.

`dispersion_method`
Bootstrap dispersion and Bayesian posterior summary: `"sd"` or `"mad"`.

`vcov` Variance-covariance matrix used to compute uncertainty estimates (e.g., for robust standard errors). This argument accepts a covariance matrix, a function which returns a covariance matrix, or a string which identifies the function to be used to compute the covariance matrix.

- A covariance matrix
- A function which returns a covariance matrix (e.g., `stats::vcov()`)
- A string which indicates the kind of uncertainty estimates to return.
  - Heteroskedasticity-consistent: `"vcovHC"`, `"HC"`, `"HC0"`, `"HC1"`, `"HC2"`, `"HC3"`, `"HC4"`, `"HC4m"`, `"HC5"`. See `?sandwich::vcovHC`
  - Cluster-robust: `"vcovCR"`, `"CR0"`, `"CR1"`, `"CR1p"`, `"CR1S"`, `"CR2"`, `"CR3"`. See `?clubSandwich::vcovCR()`
  - Bootstrap: `"vcovBS"`, `"xy"`, `"residual"`, `"wild"`, `"mammen"`, `"webb"`. See `?sandwich::vcovBS`
  - Other `sandwich` package functions: `"vcovHAC"`, `"vcovPC"`, `"vcovCL"`, `"vcovPL"`.

`vcov_args` List of arguments to be passed to the function identified by the `vcov` argument. This function is typically supplied by the **sandwich** or **clubSandwich** packages.

Please refer to their documentation (e.g., ?sandwich::vcovHAC) to see the list of available arguments. If no estimation type (argument type) is given, the default type for "HC" (or "vcovHC") equals the default from the **sandwich** package; for type "CR" (or "vcoCR"), the default is set to "CR3".

verbose
: Toggle warnings.

iterations
: For Bayesian models, this corresponds to the number of posterior draws. If NULL, will return all the draws (one for each iteration of the model). For frequentist models, if not NULL, will generate bootstrapped draws, from which bootstrapped CIs will be computed. Iterations can be accessed by running as.data.frame(..., keep_iterations = TRUE) on the output.

include_random
: If "default", include all random effects in the prediction, unless random effect variables are not in the data. If TRUE, include all random effects in the prediction (in this case, it will be checked if actually all random effect variables are in data). If FALSE, don't take them into account. Can also be a formula to specify which random effects to condition on when predicting (passed to the re.form argument). If include_random = TRUE and data is provided, make sure to include the random effect variables in data as well.

include_smooth
: For General Additive Models (GAMs). If FALSE, will fix the value of the smooth to its average, so that the predictions are not depending on it. (default), mean(), or bayestestR::map_estimate().

Details

In insight::get_predicted(), the predict argument jointly modulates two separate concepts, the **scale** and the **uncertainty interval**.

Confidence Interval (CI) vs. Prediction Interval (PI)):

- **Linear models** - lm(): For linear models, Prediction intervals (predict="prediction") show the range that likely contains the value of a new observation (in what range it is likely to fall), whereas confidence intervals (predict="expectation" or predict="link") reflect the uncertainty around the estimated parameters (and gives the range of uncertainty of the regression line). In general, Prediction Intervals (PIs) account for both the uncertainty in the model's parameters, plus the random variation of the individual values. Thus, prediction intervals are always wider than confidence intervals. Moreover, prediction intervals will not necessarily become narrower as the sample size increases (as they do not reflect only the quality of the fit, but also the variability within the data).
- **Generalized Linear models** - glm(): For binomial models, prediction intervals are somewhat useless (for instance, for a binomial (Bernoulli) model for which the dependent variable is a vector of 1s and 0s, the prediction interval is... [0, 1]).

Link scale vs. Response scale: When users set the predict argument to "expectation", the predictions are returned on the response scale, which is arguably the most convenient way to understand and visualize relationships of interest. When users set the predict argument to "link", predictions are returned on the link scale, and no transformation is applied. For instance, for a logistic regression model, the response scale corresponds to the predicted probabilities, whereas the link-scale makes predictions of log-odds (probabilities on the logit scale). Note that when users select predict="classification" in binomial models, the get_predicted() function will first calculate predictions as if the user had selected `predict="expectation"`. Then, it will round the responses in order to return the most likely outcome.

Heteroscedasticity consistent standard errors: The arguments `vcov` and `vcov_args` can be used to calculate robust standard errors for confidence intervals of predictions. These arguments, when provided in `get_predicted()`, are passed down to `get_predicted_ci()`, thus, see the related documentation there for more details.

Bayesian and Bootstrapped models and iterations: For predictions based on multiple iterations, for instance in the case of Bayesian models and bootstrapped predictions, the function used to compute the centrality (point-estimate predictions) can be modified via the `centrality_function` argument. For instance, `get_predicted(model, centrality_function = stats::median)`. The default is `mean`. Individual draws can be accessed by running `iter <- as.data.frame(get_predicted(model))`, and their iterations can be reshaped into a long format by `bayestestR::reshape_iterations(iter)`.

Value

The fitted values (i.e. predictions for the response). For Bayesian or bootstrapped models (when `iterations != NULL`), iterations (as columns and observations are rows) can be accessed via `as.data.frame()`.

See Also

`get_datagrid()`

Examples

```
data(mtcars)
x <- lm(mpg ~ cyl + hp, data = mtcars)

predictions <- get_predicted(x, ci = .95)
predictions

# Options and methods ---------------------
get_predicted(x, predict = "prediction")

# Get CI
as.data.frame(predictions)

if (require("boot")) {
  # Bootstrapped
  as.data.frame(get_predicted(x, iterations = 4))
  # Same as as.data.frame(..., keep_iterations = FALSE)
  summary(get_predicted(x, iterations = 4))
}

# Different prediction types ------------------------
data(iris)
data <- droplevels(iris[1:100, ])

# Fit a logistic model
x <- glm(Species ~ Sepal.Length, data = data, family = "binomial")
```

```
# Expectation (default): response scale + CI
pred <- get_predicted(x, predict = "expectation", ci = .95)
head(as.data.frame(pred))

# Prediction: response scale + PI
pred <- get_predicted(x, predict = "prediction", ci = .95)
head(as.data.frame(pred))

# Link: link scale + CI
pred <- get_predicted(x, predict = "link", ci = .95)
head(as.data.frame(pred))

# Classification: classification "type" + PI
pred <- get_predicted(x, predict = "classification", ci = .95)
head(as.data.frame(pred))
```

get_predicted_ci *Confidence intervals around predicted values*

Description

Confidence intervals around predicted values

Usage

```
get_predicted_ci(x, ...)

## Default S3 method:
get_predicted_ci(
  x,
  predictions = NULL,
  data = NULL,
  se = NULL,
  ci = 0.95,
  ci_type = "confidence",
  ci_method = NULL,
  dispersion_method = "sd",
  vcov = NULL,
  vcov_args = NULL,
  verbose = TRUE,
  ...
)
```

Arguments

x	A statistical model (can also be a data.frame, in which case the second argument has to be a model).

`...` Other argument to be passed, for instance to `get_predicted_ci()`.

`predictions` A vector of predicted values (as obtained by `stats::fitted()`, `stats::predict()` or `get_predicted()`).

`data` An optional data frame in which to look for variables with which to predict. If omitted, the data used to fit the model is used. Visualization matrices can be generated using `get_datagrid()`.

`se` Numeric vector of standard error of predicted values. If `NULL`, standard errors are calculated based on the variance-covariance matrix.

`ci` The interval level. Default is `NULL`, to be fast even for larger models. Set the interval level to an explicit value, e.g. `0.95`, for `95%` CI).

`ci_type` Can be `"prediction"` or `"confidence"`. Prediction intervals show the range that likely contains the value of a new observation (in what range it would fall), whereas confidence intervals reflect the uncertainty around the estimated parameters (and gives the range of the link; for instance of the regression line in a linear regressions). Prediction intervals account for both the uncertainty in the model's parameters, plus the random variation of the individual values. Thus, prediction intervals are always wider than confidence intervals. Moreover, prediction intervals will not necessarily become narrower as the sample size increases (as they do not reflect only the quality of the fit). This applies mostly for "simple" linear models (like `lm`), as for other models (e.g., `glm`), prediction intervals are somewhat useless (for instance, for a binomial model for which the dependent variable is a vector of 1s and 0s, the prediction interval is... `[0, 1]`).

`ci_method` The method for computing p values and confidence intervals. Possible values depend on model type.

- `NULL` uses the default method, which varies based on the model type.
- Most frequentist models: `"normal"` (default).
- Bayesian models: `"quantile"` (default), `"hdi"`, `"eti"`, and `"spi"`.
- Mixed effects **lme4** models: `"normal"` (default), `"satterthwaite"`, and `"kenward"`.

`dispersion_method`
Bootstrap dispersion and Bayesian posterior summary: `"sd"` or `"mad"`.

`vcov` Variance-covariance matrix used to compute uncertainty estimates (e.g., for robust standard errors). This argument accepts a covariance matrix, a function which returns a covariance matrix, or a string which identifies the function to be used to compute the covariance matrix.

- A covariance matrix
- A function which returns a covariance matrix (e.g., `stats::vcov()`)
- A string which indicates the kind of uncertainty estimates to return.
  - Heteroskedasticity-consistent: `"vcovHC"`, `"HC"`, `"HC0"`, `"HC1"`, `"HC2"`, `"HC3"`, `"HC4"`, `"HC4m"`, `"HC5"`. See `?sandwich::vcovHC`
  - Cluster-robust: `"vcovCR"`, `"CR0"`, `"CR1"`, `"CR1p"`, `"CR1S"`, `"CR2"`, `"CR3"`. See `?clubSandwich::vcovCR()`
  - Bootstrap: `"vcovBS"`, `"xy"`, `"residual"`, `"wild"`, `"mammen"`, `"webb"`. See `?sandwich::vcovBS`

- Other sandwich package functions: "vcovHAC", "vcovPC", "vcovCL", "vcovPL".

vcov_args
: List of arguments to be passed to the function identified by the vcov argument. This function is typically supplied by the **sandwich** or **clubSandwich** packages. Please refer to their documentation (e.g., ?sandwich::vcovHAC) to see the list of available arguments. If no estimation type (argument type) is given, the default type for "HC" (or "vcovHC") equals the default from the **sandwich** package; for type "CR" (or "vcoCR"), the default is set to "CR3".

verbose
: Toggle warnings.

Examples

```
# Confidence Intervals for Model Predictions
# ------------------------------------------

data(mtcars)

# Linear model
# ------------
x <- lm(mpg ~ cyl + hp, data = mtcars)
predictions <- predict(x)
ci_vals <- get_predicted_ci(x, predictions, ci_type = "prediction")
head(ci_vals)
ci_vals <- get_predicted_ci(x, predictions, ci_type = "confidence")
head(ci_vals)
ci_vals <- get_predicted_ci(x, predictions, ci = c(0.8, 0.9, 0.95))
head(ci_vals)

# Bootstrapped
# ------------
if (require("boot")) {
  predictions <- get_predicted(x, iterations = 500)
  get_predicted_ci(x, predictions)
}

if (require("datawizard") && require("bayestestR")) {
  ci_vals <- get_predicted_ci(x, predictions, ci = c(0.80, 0.95))
  head(ci_vals)
  datawizard::reshape_ci(ci_vals)

  ci_vals <- get_predicted_ci(x,
    predictions,
    dispersion_method = "MAD",
    ci_method = "HDI"
  )
  head(ci_vals)
}


# Logistic model
# --------------
x <- glm(vs ~ wt, data = mtcars, family = "binomial")
```

```
predictions <- predict(x, type = "link")
ci_vals <- get_predicted_ci(x, predictions, ci_type = "prediction")
head(ci_vals)
ci_vals <- get_predicted_ci(x, predictions, ci_type = "confidence")
head(ci_vals)
```

---

get_predictors *Get the data from model predictors*

Description

Returns the data from all predictor variables (fixed effects).

Usage

```
get_predictors(x, verbose = TRUE)
```

Arguments

x	A fitted model.
verbose	Toggle messages and warnings.

Value

The data from all predictor variables, as data frame.

Examples

```
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
head(get_predictors(m))
```

---

get_priors *Get summary of priors used for a model*

Description

Provides a summary of the prior distributions used for the parameters in a given model.

Usage

```
get_priors(x, ...)

## S3 method for class 'brmsfit'
get_priors(x, verbose = TRUE, ...)
```

Arguments

x	A Bayesian model.
...	Currently not used.
verbose	Toggle warnings and messages.

Value

A data frame with a summary of the prior distributions used for the parameters in a given model.

Examples

```
## Not run:
library(rstanarm)
model <- stan_glm(Sepal.Width ~ Species * Petal.Length, data = iris)
get_priors(model)

## End(Not run)
```

---

get_random — *Get the data from random effects*

Description

Returns the data from all random effects terms.

Usage

```
get_random(x)
```

Arguments

x	A fitted mixed model.

Value

The data from all random effects terms, as data frame. Or `NULL` if model has no random effects.

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  # prepare some data...
  sleepstudy$mygrp <- sample(1:5, size = 180, replace = TRUE)
  sleepstudy$mysubgrp <- NA
  for (i in 1:5) {
    filter_group <- sleepstudy$mygrp == i
    sleepstudy$mysubgrp[filter_group] <-
```

```
    sample(1:30, size = sum(filter_group), replace = TRUE)
  }

  m <- lmer(
    Reaction ~ Days + (1 | mygrp / mysubgrp) + (1 | Subject),
    data = sleepstudy
  )

  head(get_random(m))
}
```

`get_residuals` *Extract model residuals*

Description

Returns the residuals from regression models.

Usage

```
get_residuals(x, ...)

## Default S3 method:
get_residuals(x, weighted = FALSE, verbose = TRUE, ...)
```

Arguments

`x`	A model.
`...`	Passed down to `residuals()`, if possible.
`weighted`	Logical, if `TRUE`, returns weighted residuals.
`verbose`	Toggle warnings and messages.

Value

The residuals, or `NULL` if this information could not be accessed.

Note

This function returns the default type of residuals, i.e. for the response from linear models, the deviance residuals for models of class `glm` etc. To access different types, pass down the `type` argument (see 'Examples').

This function is a robust alternative to `residuals()`, as it works for some special model objects that otherwise do not respond properly to calling `residuals()`.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_residuals(m)

m <- glm(vs ~ wt + cyl + mpg, data = mtcars, family = binomial())
get_residuals(m) # type = "deviance" by default
get_residuals(m, type = "response")
```

---

get_response   *Get the values from the response variable*

Description

Returns the values the response variable(s) from a model object. If the model is a multivariate response model, a data frame with values from all response variables is returned.

Usage

```
get_response(x, select = NULL, verbose = TRUE)
```

Arguments

x	A fitted model.
select	Optional name(s) of response variables for which to extract values. Can be used in case of regression models with multiple response variables.
verbose	Toggle warnings.

Value

The values of the response variable, as vector, or a data frame if x has more than one defined response variable.

Examples

```
if (require("lme4")) {
  data(cbpp)
  cbpp$trials <- cbpp$size - cbpp$incidence

  m <- glm(cbind(incidence, trials) ~ period, data = cbpp, family = binomial)
  head(get_response(m))
  get_response(m, select = "incidence")
}

data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_response(m)
```

`get_sigma` *Get residual standard deviation from models*

Description

Returns `sigma`, which corresponds the estimated standard deviation of the residuals. This function extends the `sigma()` base R generic for models that don't have implemented it. It also computes the confidence interval (CI), which is stored as an attribute.

Sigma is a key-component of regression models, and part of the so-called auxiliary parameters that are estimated. Indeed, linear models for instance assume that the residuals comes from a normal distribution with mean 0 and standard deviation `sigma`. See the details section below for more information about its interpretation and calculation.

Usage

```
get_sigma(x, ci = NULL, verbose = TRUE)
```

Arguments

`x`	A model.
`ci`	Scalar, the CI level. The default (`NULL`) returns no CI.
`verbose`	Toggle messages and warnings.

Details

Interpretation of Sigma: The residual standard deviation, σ, indicates that the predicted outcome will be within +/- σ units of the linear predictor for approximately `68%` of the data points (*Gelman, Hill & Vehtari 2020, p.84*). In other words, the residual standard deviation indicates the accuracy for a model to predict scores, thus it can be thought of as "a measure of the average distance each observation falls from its prediction from the model" (*Gelman, Hill & Vehtari 2020, p.168*). σ can be considered as a measure of the unexplained variation in the data, or of the precision of inferences about regression coefficients.

Calculation of Sigma: By default, `get_sigma()` tries to extract sigma by calling `stats::sigma()`. If the model-object has no `sigma()` method, the next step is calculating sigma as square-root of the model-deviance divided by the residual degrees of freedom. Finally, if even this approach fails, and `x` is a mixed model, the residual standard deviation is accessed using the square-root from `get_variance_residual()`.

Value

The residual standard deviation (sigma), or `NULL` if this information could not be accessed.

References

Gelman, A., Hill, J., & Vehtari, A. (2020). Regression and Other Stories. Cambridge University Press.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_sigma(m)
```

---

get_statistic *Get statistic associated with estimates*

Description

Returns the statistic (*t*, z, ...) for model estimates. In most cases, this is the related column from `coef(summary())`.

Usage

```
get_statistic(x, ...)

## Default S3 method:
get_statistic(x, column_index = 3, verbose = TRUE, ...)

## S3 method for class 'glmmTMB'
get_statistic(
  x,
  component = c("all", "conditional", "zi", "zero_inflated", "dispersion"),
  ...
)

## S3 method for class 'clm2'
get_statistic(x, component = c("all", "conditional", "scale"), ...)

## S3 method for class 'betamfx'
get_statistic(
  x,
  component = c("all", "conditional", "precision", "marginal"),
  ...
)

## S3 method for class 'logitmfx'
get_statistic(x, component = c("all", "conditional", "marginal"), ...)

## S3 method for class 'mjoint'
get_statistic(x, component = c("all", "conditional", "survival"), ...)

## S3 method for class 'emmGrid'
get_statistic(x, ci = 0.95, adjust = "none", merge_parameters = FALSE, ...)

## S3 method for class 'gee'
```

```
get_statistic(x, robust = FALSE, ...)

## S3 method for class 'betareg'
get_statistic(x, component = c("all", "conditional", "precision"), ...)

## S3 method for class 'DirichletRegModel'
get_statistic(x, component = c("all", "conditional", "precision"), ...)
```

Arguments

- `x` A model.
- `...` Currently not used.
- `column_index` For model objects that have no defined `get_statistic()` method yet, the default method is called. This method tries to extract the statistic column from `coef(summary())`, where the index of the column that is being pulled is `column_index`. Defaults to 3, which is the default statistic column for most models' summary-output.
- `verbose` Toggle messages and warnings.
- `component` Should all parameters, parameters for the conditional model, or for the zero-inflated part of the model be returned? Applies to models with zero-inflated component. `component` may be one of `"conditional"`, `"zi"`, `"zero-inflated"` or `"all"` (default). For models with smooth terms, `component = "smooth_terms"` is also possible. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.
- `ci` Confidence Interval (CI) level. Default to `0.95` (`95%`). Currently only applies to objects of class `emmGrid`.
- `adjust` Character value naming the method used to adjust p-values or confidence intervals. See `?emmeans::summary.emmGrid` for details.
- `merge_parameters` Logical, if `TRUE` and `x` has multiple columns for parameter names (like `emmGrid` objects may have), these are merged into a single parameter column, with parameters names and values as values.
- `robust` Logical, if `TRUE`, test statistic based on robust standard errors is returned.

Value

A data frame with the model's parameter names and the related test statistic.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_statistic(m)
```

get_transformation *Return function of transformed response variables*

Description

This functions checks whether any transformation, such as log- or exp-transforming, was applied to the response variable (dependent variable) in a regression formula, and returns the related function that was used for transformation.

Usage

```
get_transformation(x)
```

Arguments

x	A regression model.

Value

A list of two functions: `$transformation`, the function that was used to transform the response variable; `$inverse`, the inverse-function of `$transformation` (can be used for "back-transformation"). If no transformation was applied, both list-elements `$transformation` and `$inverse` just return `function(x) x`. If transformation is unknown, `NULL` is returned.

Examples

```
# identity, no transformation
model <- lm(Sepal.Length ~ Species, data = iris)
get_transformation(model)

# log-transformation
model <- lm(log(Sepal.Length) ~ Species, data = iris)
get_transformation(model)

# log-function
get_transformation(model)$transformation(.3)
log(.3)

# inverse function is exp()
get_transformation(model)$inverse(.3)
exp(.3)
```

get_varcov *Get variance-covariance matrix from models*

Description

Returns the variance-covariance, as retrieved by `stats::vcov()`, but works for more model objects that probably don't provide a `vcov()`-method.

Usage

```
get_varcov(x, ...)

## Default S3 method:
get_varcov(x, verbose = TRUE, vcov = NULL, vcov_args = NULL, ...)

## S3 method for class 'betareg'
get_varcov(
  x,
  component = c("conditional", "precision", "all"),
  verbose = TRUE,
  ...
)

## S3 method for class 'clm2'
get_varcov(x, component = c("all", "conditional", "scale"), ...)

## S3 method for class 'truncreg'
get_varcov(x, component = c("conditional", "all"), verbose = TRUE, ...)

## S3 method for class 'hurdle'
get_varcov(
  x,
  component = c("conditional", "zero_inflated", "zi", "all"),
  vcov = NULL,
  vcov_args = NULL,
  verbose = TRUE,
  ...
)

## S3 method for class 'glmmTMB'
get_varcov(
  x,
  component = c("conditional", "zero_inflated", "zi", "dispersion", "all"),
  verbose = TRUE,
  ...
)
```

```
## S3 method for class 'MixMod'
get_varcov(
  x,
  effects = c("fixed", "random"),
 component = c("conditional", "zero_inflated", "zi", "dispersion", "auxiliary", "all"),
  verbose = TRUE,
  ...
)

## S3 method for class 'brmsfit'
get_varcov(x, component = "conditional", verbose = TRUE, ...)

## S3 method for class 'betamfx'
get_varcov(
  x,
  component = c("conditional", "precision", "all"),
  verbose = TRUE,
  ...
)

## S3 method for class 'aov'
get_varcov(x, complete = FALSE, verbose = TRUE, ...)

## S3 method for class 'mixor'
get_varcov(x, effects = c("all", "fixed", "random"), verbose = TRUE, ...)
```

Arguments

`x` A model.

`...` Currently not used.

`verbose` Toggle warnings.

`vcov` Variance-covariance matrix used to compute uncertainty estimates (e.g., for robust standard errors). This argument accepts a covariance matrix, a function which returns a covariance matrix, or a string which identifies the function to be used to compute the covariance matrix.

- A covariance matrix
- A function which returns a covariance matrix (e.g., `stats::vcov()`)
- A string which indicates the kind of uncertainty estimates to return.
  - Heteroskedasticity-consistent: `"vcovHC"`, `"HC"`, `"HC0"`, `"HC1"`, `"HC2"`, `"HC3"`, `"HC4"`, `"HC4m"`, `"HC5"`. See `?sandwich::vcovHC`
  - Cluster-robust: `"vcovCR"`, `"CR0"`, `"CR1"`, `"CR1p"`, `"CR1S"`, `"CR2"`, `"CR3"`. See `?clubSandwich::vcovCR()`
  - Bootstrap: `"vcovBS"`, `"xy"`, `"residual"`, `"wild"`, `"mammen"`, `"webb"`. See `?sandwich::vcovBS`
  - Other `sandwich` package functions: `"vcovHAC"`, `"vcovPC"`, `"vcovCL"`, `"vcovPL"`.

`vcov_args` List of arguments to be passed to the function identified by the `vcov` argument. This function is typically supplied by the **sandwich** or **clubSandwich** packages. Please refer to their documentation (e.g., `?sandwich::vcovHAC`) to see the list of available arguments. If no estimation type (argument `type`) is given, the default type for `"HC"` (or `"vcovHC"`) equals the default from the **sandwich** package; for type `"CR"` (or `"vcoCR"`), the default is set to `"CR3"`.

`component` Should the complete variance-covariance matrix of the model be returned, or only for specific model components only (like count or zero-inflated model parts)? Applies to models with zero-inflated component, or models with precision (e.g. `betareg`) component. `component` may be one of `"conditional"`, `"zi"`, `"zero-inflated"`, `"dispersion"`, `"precision"`, or `"all"`. May be abbreviated. Note that the *conditional* component is also called *count* or *mean* component, depending on the model.

`effects` Should the complete variance-covariance matrix of the model be returned, or only for specific model parameters only? Currently only applies to models of class `mixor`.

`complete` Logical, if `TRUE`, for `aov`, returns the full variance-covariance matrix.

Value

The variance-covariance matrix, as `matrix`-object.

Note

`get_varcov()` tries to return the nearest positive definite matrix in case of negative eigenvalues of the variance-covariance matrix. This ensures that it is still possible, for instance, to calculate standard errors of model parameters. A message is shown when the matrix is negative definite and a corrected matrix is returned.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
get_varcov(m)

# vcov of zero-inflation component from hurdle-model
if (require("pscl")) {
  data("bioChemists", package = "pscl")
  mod <- hurdle(art ~ phd + fem | ment, data = bioChemists, dist = "negbin")
  get_varcov(mod, component = "zero_inflated")
}

# robust vcov of, count component from hurdle-model
if (require("pscl") && require("sandwich")) {
  data("bioChemists", package = "pscl")
  mod <- hurdle(art ~ phd + fem | ment, data = bioChemists, dist = "negbin")
  get_varcov(
    mod,
    component = "conditional",
    vcov = "BS",
```

```
    vcov_args = list(R = 50)
  )
}
```

get_variance *Get variance components from random effects models*

Description

This function extracts the different variance components of a mixed model and returns the result as list. Functions like `get_variance_residual(x)` or `get_variance_fixed(x)` are shortcuts for `get_variance(x, component = "residual")` etc.

Usage

```
get_variance(
  x,
  component = c("all", "fixed", "random", "residual", "distribution", "dispersion",
    "intercept", "slope", "rho01", "rho00"),
  verbose = TRUE,
  ...
)

get_variance_residual(x, verbose = TRUE, ...)

get_variance_fixed(x, verbose = TRUE, ...)

get_variance_random(x, verbose = TRUE, tolerance = 1e-05, ...)

get_variance_distribution(x, verbose = TRUE, ...)

get_variance_dispersion(x, verbose = TRUE, ...)

get_variance_intercept(x, verbose = TRUE, ...)

get_variance_slope(x, verbose = TRUE, ...)

get_correlation_slope_intercept(x, verbose = TRUE, ...)

get_correlation_slopes(x, verbose = TRUE, ...)
```

Arguments

x	A mixed effects model.
component	Character value, indicating the variance component that should be returned. By default, all variance components are returned. The distribution-specific ("distribution") and residual ("residual") variance are the most computational intensive components, and hence may take a few seconds to calculate.

`verbose` Toggle off warnings.

`...` Currently not used.

`tolerance` Tolerance for singularity check of random effects, to decide whether to compute random effect variances or not. Indicates up to which value the convergence result is accepted. The larger tolerance is, the stricter the test will be. See `performance::check_singularity()`.

Details

This function returns different variance components from mixed models, which are needed, for instance, to calculate r-squared measures or the intraclass-correlation coefficient (ICC).

Fixed effects variance: The fixed effects variance, σ_f^2, is the variance of the matrix-multiplication $\beta * X$ (parameter vector by model matrix).

Random effects variance: The random effect variance, σ_i^2, represents the *mean* random effect variance of the model. Since this variance reflect the "average" random effects variance for mixed models, it is also appropriate for models with more complex random effects structures, like random slopes or nested random effects. Details can be found in *Johnson 2014*, in particular equation 10. For simple random-intercept models, the random effects variance equals the random-intercept variance.

Distribution-specific variance: The distribution-specific variance, σ_d^2, depends on the model family. For Gaussian models, it is σ^2 (i.e. `sigma(model)^2`). For models with binary outcome, it is $\pi^2/3$ for logit-link, `1` for probit-link, and $\pi^2/6$ for cloglog-links. Models from Gamma-families use μ^2 (as obtained from `family$variance()`). For all other models, the distribution-specific variance is based on lognormal approximation, $log(1 + var(x)/\mu^2)$ (see *Nakagawa et al. 2017*). The expected variance of a zero-inflated model is computed according to *Zuur et al. 2012, p277*.

Variance for the additive overdispersion term: The variance for the additive overdispersion term, σ_e^2, represents “the excess variation relative to what is expected from a certain distribution” (Nakagawa et al. 2017). In (most? many?) cases, this will be `0`.

Residual variance: The residual variance, σ_ϵ^2, is simply $\sigma_d^2 + \sigma_e^2$.

Random intercept variance: The random intercept variance, or *between-subject* variance (τ_{00}), is obtained from `VarCorr()`. It indicates how much groups or subjects differ from each other, while the residual variance σ_ϵ^2 indicates the *within-subject variance*.

Random slope variance: The random slope variance (τ_{11}) is obtained from `VarCorr()`. This measure is only available for mixed models with random slopes.

Random slope-intercept correlation: The random slope-intercept correlation (ρ_{01}) is obtained from `VarCorr()`. This measure is only available for mixed models with random intercepts and slopes.

Value

A list with following elements:

- `var.fixed`, variance attributable to the fixed effects
- `var.random`, (mean) variance of random effects
- `var.residual`, residual variance (sum of dispersion and distribution)
- `var.distribution`, distribution-specific variance
- `var.dispersion`, variance due to additive dispersion
- `var.intercept`, the random-intercept-variance, or between-subject-variance (τ_{00})
- `var.slope`, the random-slope-variance (τ_{11})
- `cor.slope_intercept`, the random-slope-intercept-correlation (ρ_{01})
- `cor.slopes`, the correlation between random slopes (ρ_{00})

Note

This function supports models of class `merMod` (including models from **blme**), `clmm`, `cpglmm`, `glmmadmb`, `glmmTMB`, `MixMod`, `lme`, `mixed`, `rlmerMod`, `stanreg`, `brmsfit` or `wbm`. Support for objects of class `MixMod` (**GLMMadaptive**), `lme` (**nlme**) or `brmsfit` (**brms**) is experimental and may not work for all models.

References

- Johnson, P. C. D. (2014). Extension of Nakagawa & Schielzeth's R2 GLMM to random slopes models. Methods in Ecology and Evolution, 5(9), 944–946. doi:10.1111/2041210X.12225
- Nakagawa, S., Johnson, P. C. D., & Schielzeth, H. (2017). The coefficient of determination R2 and intra-class correlation coefficient from generalized linear mixed-effects models revisited and expanded. Journal of The Royal Society Interface, 14(134), 20170213. doi:10.1098/rsif.2017.0213
- Zuur, A. F., Savel'ev, A. A., & Ieno, E. N. (2012). Zero inflated models and generalized linear mixed models with R. Newburgh, United Kingdom: Highland Statistics.

Examples

```
## Not run:
library(lme4)
data(sleepstudy)
m <- lmer(Reaction ~ Days + (1 + Days | Subject), data = sleepstudy)

get_variance(m)
get_variance_fixed(m)
get_variance_residual(m)

## End(Not run)
```

get_weights — *Get the values from model weights*

Description

Returns weighting variable of a model.

Usage

```
get_weights(x, ...)

## Default S3 method:
get_weights(x, na_rm = FALSE, null_as_ones = FALSE, ...)
```

Arguments

x	A fitted model.
...	Currently not used.
na_rm	Logical, if TRUE, removes possible missing values.
null_as_ones	Logical, if TRUE, will return a vector of 1 if no weights were specified in the model (as if the weights were all set to 1).

Value

The weighting variable, or NULL if no weights were specified or if weights were 1. If the weighting variable should also be returned (instead of NULL), when all weights are set to 1 (i.e. no weighting), set null_as_ones = TRUE.

Examples

```
data(mtcars)
set.seed(123)
mtcars$weight <- rnorm(nrow(mtcars), 1, .3)

# LMs
m <- lm(mpg ~ wt + cyl + vs, data = mtcars, weights = weight)
get_weights(m)

get_weights(lm(mpg ~ wt, data = mtcars), null_as_ones = TRUE)

# GLMs
m <- glm(vs ~ disp + mpg, data = mtcars, weights = weight, family = quasibinomial)
get_weights(m)
m <- glm(cbind(cyl, gear) ~ mpg, data = mtcars, weights = weight, family = binomial)
get_weights(m)
```

has_intercept — *Checks if model has an intercept*

Description

Checks if model has an intercept.

Usage

```
has_intercept(x, verbose = TRUE)
```

Arguments

`x`	A model object.
`verbose`	Toggle warnings.

Value

`TRUE` if `x` has an intercept, `FALSE` otherwise.

Examples

```
model <- lm(mpg ~ 0 + gear, data = mtcars)
has_intercept(model)

model <- lm(mpg ~ gear, data = mtcars)
has_intercept(model)

if (require("lme4")) {
  model <- lmer(Reaction ~ 0 + Days + (Days | Subject), data = sleepstudy)
  has_intercept(model)

  model <- lmer(Reaction ~ Days + (Days | Subject), data = sleepstudy)
  has_intercept(model)
}
```

is_converged — *Convergence test for mixed effects models*

Description

`is_converged()` provides an alternative convergence test for `merMod`-objects.

Usage

```
is_converged(x, tolerance = 0.001, ...)
```

Arguments

`x`	A `merMod` or `glmmTMB`-object.
`tolerance`	Indicates up to which value the convergence result is accepted. The smaller `tolerance` is, the stricter the test will be.
`...`	Currently not used.

Details

Convergence and log-likelihood: Convergence problems typically arise when the model hasn't converged to a solution where the log-likelihood has a true maximum. This may result in unreliable and overly complex (or non-estimable) estimates and standard errors.

Inspect model convergence: **lme4** performs a convergence-check (see `?lme4::convergence`), however, as as discussed here and suggested by one of the lme4-authors in this comment, this check can be too strict. `is_converged()` thus provides an alternative convergence test for `merMod`-objects.

Resolving convergence issues: Convergence issues are not easy to diagnose. The help page on `?lme4::convergence` provides most of the current advice about how to resolve convergence issues. Another clue might be large parameter values, e.g. estimates (on the scale of the linear predictor) larger than 10 in (non-identity link) generalized linear model *might* indicate complete separation. Complete separation can be addressed by regularization, e.g. penalized regression or Bayesian regression with appropriate priors on the fixed effects.

Convergence versus Singularity: Note the different meaning between singularity and convergence: singularity indicates an issue with the "true" best estimate, i.e. whether the maximum likelihood estimation for the variance-covariance matrix of the random effects is positive definite or only semi-definite. Convergence is a question of whether we can assume that the numerical optimization has worked correctly or not.

Value

`TRUE` if convergence is fine and `FALSE` if convergence is suspicious. Additionally, the convergence value is returned as attribute.

Examples

```
if (require("lme4")) {
  data(cbpp)
  set.seed(1)
  cbpp$x <- rnorm(nrow(cbpp))
  cbpp$x2 <- runif(nrow(cbpp))

  model <- glmer(
    cbind(incidence, size - incidence) ~ period + x + x2 + (1 + x | herd),
    data = cbpp,
    family = binomial()
  )

  is_converged(model)
```

```
}

## Not run:
if (require("glmmTMB")) {
  model <- glmmTMB(Sepal.Length ~ poly(Petal.Width, 4) * poly(Petal.Length, 4) +
  (1+poly(Petal.Width, 4)|Species), data=iris)

  is_converged(model)
}

## End(Not run)
```

`is_empty_object` *Check if object is empty*

Description

Check if object is empty

Usage

```
is_empty_object(x)
```

Arguments

x	A list, a vector, or a dataframe.

Value

A logical indicating whether the entered object is empty.

Examples

```
is_empty_object(c(1, 2, 3, NA))
is_empty_object(list(NULL, c(NA, NA)))
is_empty_object(list(NULL, NA))
```

`is_gam_model` *Checks if a model is a generalized additive model*

Description

Small helper that checks if a model is a generalized additive model.

Usage

```
is_gam_model(x)
```

Arguments

x A model object.

Value

A logical, TRUE if x is a generalized additive model *and* has smooth-terms

Note

This function only returns TRUE when the model inherits from a typical GAM model class *and* when smooth terms are present in the model formula. If model has no smooth terms or is not from a typical gam class, FALSE is returned.

Examples

```
if (require("mgcv")) {
  data(iris)
  model1 <- lm(Petal.Length ~ Petal.Width + Sepal.Length, data = iris)
  model2 <- gam(Petal.Length ~ Petal.Width + s(Sepal.Length), data = iris)
  is_gam_model(model1)
  is_gam_model(model2)
}
```

---

is_mixed_model *Checks if a model is a mixed effects model*

Description

Small helper that checks if a model is a mixed effects model, i.e. if it the model has random effects.

Usage

```
is_mixed_model(x)
```

Arguments

x A model object.

Value

A logical, TRUE if x is a mixed model.

Examples

```
data(mtcars)
model <- lm(mpg ~ wt + cyl + vs, data = mtcars)
is_mixed_model(model)

if (require("lme4")) {
  data(sleepstudy)
  model <- lmer(Reaction ~ Days + (1 | Subject), data = sleepstudy)
  is_mixed_model(model)
}
```

`is_model` *Checks if an object is a regression model or statistical test object*

Description

Small helper that checks if a model is a regression model or a statistical object. `is_regression_model()` is stricter and only returns `TRUE` for regression models, but not for, e.g., `htest` objects.

Usage

```
is_model(x)

is_regression_model(x)
```

Arguments

`x`	An object.

Details

This function returns `TRUE` if `x` is a model object.

Value

A logical, `TRUE` if `x` is a (supported) model object.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)

is_model(m)
is_model(mtcars)

test <- t.test(1:10, y = c(7:20))
is_model(test)
is_regression_model(test)
```

`is_model_supported` *Checks if a regression model object is supported in* **insight** *package*

Description

Small helper that checks if a model is a *supported* (regression) model object. `supported_models()` prints a list of currently supported model classes.

Usage

```
is_model_supported(x)

supported_models()
```

Arguments

`x`	An object.

Details

This function returns `TRUE` if `x` is a model object that works with the package's functions. A list of supported models can also be found here: https://github.com/easystats/insight.

Value

A logical, `TRUE` if `x` is a (supported) model object.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)

is_model_supported(m)
is_model_supported(mtcars)

# to see all supported models
supported_models()
```

`is_multivariate` *Checks if an object stems from a multivariate response model*

Description

Small helper that checks if a model is a multivariate response model, i.e. a model with multiple outcomes.

Usage

```
is_multivariate(x)
```

Arguments

`x`	A model object, or an object returned by a function from this package.

Value

A logical, `TRUE` if either `x` is a model object and is a multivariate response model, or `TRUE` if a return value from a function of **insight** is from a multivariate response model.

Examples

```
## Not run:
library(rstanarm)
data("pbcLong")
model <- stan_mvmer(
  formula = list(
    logBili ~ year + (1 | id),
    albumin ~ sex + year + (year | id)
  ),
  data = pbcLong,
  chains = 1, cores = 1, seed = 12345, iter = 1000
)

f <- find_formula(model)
is_multivariate(model)
is_multivariate(f)

## End(Not run)
```

is_nested_models — *Checks whether a list of models are nested models*

Description

Checks whether a list of models are nested models, strictly following the order they were passed to the function.

Usage

```
is_nested_models(...)
```

Arguments

`...`	Multiple regression model objects.

Value

`TRUE` if models are nested, `FALSE` otherwise. If models are nested, also returns two attributes that indicate whether nesting of models is in decreasing or increasing order.

Examples

```
m1 <- lm(Sepal.Length ~ Petal.Width + Species, data = iris)
m2 <- lm(Sepal.Length ~ Species, data = iris)
m3 <- lm(Sepal.Length ~ Petal.Width, data = iris)
m4 <- lm(Sepal.Length ~ 1, data = iris)

is_nested_models(m1, m2, m4)
is_nested_models(m4, m2, m1)
is_nested_models(m1, m2, m3)
```

is_nullmodel — *Checks if model is a null-model (intercept-only)*

Description

Checks if model is a null-model (intercept-only), i.e. if the conditional part of the model has no predictors.

Usage

```
is_nullmodel(x)
```

Arguments

`x`	A model object.

Value

TRUE if x is a null-model, FALSE otherwise.

Examples

```
model <- lm(mpg ~ 1, data = mtcars)
is_nullmodel(model)

model <- lm(mpg ~ gear, data = mtcars)
is_nullmodel(model)

if (require("lme4")) {
  model <- lmer(Reaction ~ 1 + (Days | Subject), data = sleepstudy)
  is_nullmodel(model)

  model <- lmer(Reaction ~ Days + (Days | Subject), data = sleepstudy)
  is_nullmodel(model)
}
```

---

link_function   *Get link-function from model object*

Description

Returns the link-function from a model object.

Usage

```
link_function(x, ...)

## S3 method for class 'betamfx'
link_function(x, what = c("mean", "precision"), ...)

## S3 method for class 'gamlss'
link_function(x, what = c("mu", "sigma", "nu", "tau"), ...)

## S3 method for class 'betareg'
link_function(x, what = c("mean", "precision"), ...)

## S3 method for class 'DirichletRegModel'
link_function(x, what = c("mean", "precision"), ...)
```

Arguments

x	A fitted model.
...	Currently not used.
what	For gamlss models, indicates for which distribution parameter the link (inverse) function should be returned; for betareg or DirichletRegModel, can be "mean" or "precision".

Value

A function, describing the link-function from a model-object. For multivariate-response models, a list of functions is returned.

Examples

```
# example from ?stats::glm
counts <- c(18, 17, 15, 20, 10, 20, 25, 13, 12)
outcome <- gl(3, 1, 9)
treatment <- gl(3, 3)
m <- glm(counts ~ outcome + treatment, family = poisson())

link_function(m)(.3)
# same as
log(.3)
```

---

link_inverse *Get link-inverse function from model object*

Description

Returns the link-inverse function from a model object.

Usage

```
link_inverse(x, ...)

## S3 method for class 'betareg'
link_inverse(x, what = c("mean", "precision"), ...)

## S3 method for class 'DirichletRegModel'
link_inverse(x, what = c("mean", "precision"), ...)

## S3 method for class 'betamfx'
link_inverse(x, what = c("mean", "precision"), ...)

## S3 method for class 'gamlss'
link_inverse(x, what = c("mu", "sigma", "nu", "tau"), ...)
```

Arguments

x	A fitted model.
...	Currently not used.
what	For gamlss models, indicates for which distribution parameter the link (inverse) function should be returned; for betareg or DirichletRegModel, can be "mean" or "precision".

Value

A function, describing the inverse-link function from a model-object. For multivariate-response models, a list of functions is returned.

Examples

```
# example from ?stats::glm
counts <- c(18, 17, 15, 20, 10, 20, 25, 13, 12)
outcome <- gl(3, 1, 9)
treatment <- gl(3, 3)
m <- glm(counts ~ outcome + treatment, family = poisson())

link_inverse(m)(.3)
# same as
exp(.3)
```

`model_info` *Access information from model objects*

Description

Retrieve information from model objects.

Usage

```
model_info(x, ...)

## Default S3 method:
model_info(x, verbose = TRUE, ...)
```

Arguments

`x`	A fitted model.
`...`	Currently not used.
`verbose`	Toggle off warnings.

Details

`model_info()` returns a list with information about the model for many different model objects. Following information is returned, where all values starting with `is_` are logicals.

- `is_binomial`: family is binomial (but not negative binomial)
- `is_bernoulli`: special case of binomial models: family is Bernoulli
- `is_poisson`: family is poisson
- `is_negbin`: family is negative binomial
- `is_count`: model is a count model (i.e. family is either poisson or negative binomial)

- `is_beta`: family is beta
- `is_betabinomial`: family is beta-binomial
- `is_dirichlet`: family is dirichlet
- `is_exponential`: family is exponential (e.g. Gamma or Weibull)
- `is_logit`: model has logit link
- `is_probit`: model has probit link
- `is_linear`: family is gaussian
- `is_tweedie`: family is tweedie
- `is_ordinal`: family is ordinal or cumulative link
- `is_cumulative`: family is ordinal or cumulative link
- `is_multinomial`: family is multinomial or categorical link
- `is_categorical`: family is categorical link
- `is_censored`: model is a censored model (has a censored response, including survival models)
- `is_truncated`: model is a truncated model (has a truncated response)
- `is_survival`: model is a survival model
- `is_zero_inflated`: model has zero-inflation component
- `is_hurdle`: model has zero-inflation component and is a hurdle-model (truncated family distribution)
- `is_dispersion`: model has dispersion component (not only dispersion *parameter*)
- `is_mixed`: model is a mixed effects model (with random effects)
- `is_multivariate`: model is a multivariate response model (currently only works for *brmsfit* objects)
- `is_trial`: model response contains additional information about the trials
- `is_bayesian`: model is a Bayesian model
- `is_gam`: model is a generalized additive model
- `is_anova`: model is an Anova object
- `is_ttest`: model is an an object of class `htest`, returned by `t.test()`
- `is_correlation`: model is an an object of class `htest`, returned by `cor.test()`
- `is_ranktest`: model is an an object of class `htest`, returned by `cor.test()` (if Spearman's rank correlation), `wilcox.text()` or `kruskal.test()`.
- `is_levenetest`: model is an an object of class `anova`, returned by `car::leveneTest()`.
- `is_onewaytest`: model is an an object of class `htest`, returned by `oneway.test()`
- `is_proptest`: model is an an object of class `htest`, returned by `prop.test()`
- `is_binomtest`: model is an an object of class `htest`, returned by `binom.test()`
- `is_chi2test`: model is an an object of class `htest`, returned by `chisq.test()`
- `is_xtab`: model is an an object of class `htest` or `BFBayesFactor`, and test-statistic stems from a contingency table (i.e. `chisq.test()` or `BayesFactor::contingencyTableBF()`).
- `link_function`: the link-function
- `family`: the family-object
- `n_obs`: number of observations

Value

A list with information about the model, like family, link-function etc. (see 'Details').

Examples

```
ldose <- rep(0:5, 2)
numdead <- c(1, 4, 9, 13, 18, 20, 0, 2, 6, 10, 12, 16)
sex <- factor(rep(c("M", "F"), c(6, 6)))
SF <- cbind(numdead, numalive = 20 - numdead)
dat <- data.frame(ldose, sex, SF, stringsAsFactors = FALSE)
m <- glm(SF ~ sex * ldose, family = binomial)

model_info(m)
## Not run:
library(glmmTMB)
data("Salamanders")
m <- glmmTMB(
  count ~ spp + cover + mined + (1 | site),
  ziformula = ~ spp + mined,
  dispformula = ~DOY,
  data = Salamanders,
  family = nbinom2
)

## End(Not run)

model_info(m)
```

`model_name` *Name the model*

Description

Returns the "name" (class attribute) of a model, possibly including further information.

Usage

```
model_name(x, ...)

## Default S3 method:
model_name(x, include_formula = FALSE, include_call = FALSE, ...)
```

Arguments

`x`	A model.
`...`	Currently not used.
`include_formula`	Should the name include the model's formula.
`include_call`	If `TRUE`, will return the function call as a name.

Value

A character string of a name (which usually equals the model's class attribute).

Examples

```
m <- lm(Sepal.Length ~ Petal.Width, data = iris)
model_name(m)
model_name(m, include_formula = TRUE)
model_name(m, include_call = TRUE)

if (require("lme4")) {
  model_name(lmer(Sepal.Length ~ Sepal.Width + (1 | Species), data = iris))
}
```

`null_model` *Compute intercept-only model for regression models*

Description

This function computes the null-model (i.e. `(y ~ 1)`) of a model. For mixed models, the null-model takes random effects into account.

Usage

```
null_model(model, verbose = TRUE, ...)
```

Arguments

`model`	A (mixed effects) model.
`verbose`	Toggle off warnings.
`...`	Arguments passed to or from other methods.

Value

The null-model of `x`

Examples

```
if (require("lme4")) {
  data(sleepstudy)
  m <- lmer(Reaction ~ Days + (1 + Days | Subject), data = sleepstudy)
  summary(m)
  summary(null_model(m))
}
```

n_obs — *Get number of observations from a model*

Description

This method returns the number of observation that were used to fit the model, as numeric value.

Usage

```
n_obs(x, ...)

## S3 method for class 'glm'
n_obs(x, disaggregate = FALSE, ...)

## S3 method for class 'svyolr'
n_obs(x, weighted = FALSE, ...)

## S3 method for class 'afex_aov'
n_obs(x, shape = c("long", "wide"), ...)

## S3 method for class 'stanmvreg'
n_obs(x, select = NULL, ...)
```

Arguments

x	A fitted model.
...	Currently not used.
disaggregate	For binomial models with aggregated data, `n_obs()` returns the number of data rows by default. If `disaggregate = TRUE`, the total number of trials is returned instead (determined by summing the results of `weights()` for aggregated data, which will be either the weights input for proportion success response or the row sums of the response matrix if matrix response, see 'Examples').
weighted	For survey designs, returns the weighted sample size.
shape	Return long or wide data? Only applicable in repeated measures designs.
select	Optional name(s) of response variables for which to extract values. Can be used in case of regression models with multiple response variables.

Value

The number of observations used to fit the model, or `NULL` if this information is not available.

Examples

```
data(mtcars)
m <- lm(mpg ~ wt + cyl + vs, data = mtcars)
n_obs(m)
```

```
if (require("lme4")) {
  data(cbpp, package = "lme4")
  m <- glm(
    cbind(incidence, size - incidence) ~ period,
    data = cbpp,
    family = binomial(link = "logit")
  )
  n_obs(m)
  n_obs(m, disaggregate = TRUE)
}
```

`n_parameters` *Count number of parameters in a model*

Description

Returns the number of parameters (coefficients) of a model.

Usage

```
n_parameters(x, ...)

## Default S3 method:
n_parameters(x, remove_nonestimable = FALSE, ...)

## S3 method for class 'merMod'
n_parameters(
  x,
  effects = c("fixed", "random"),
  remove_nonestimable = FALSE,
  ...
)

## S3 method for class 'glmmTMB'
n_parameters(
  x,
  effects = c("fixed", "random"),
  component = c("all", "conditional", "zi", "zero_inflated"),
  remove_nonestimable = FALSE,
  ...
)

## S3 method for class 'zeroinfl'
n_parameters(
  x,
  component = c("all", "conditional", "zi", "zero_inflated"),
  remove_nonestimable = FALSE,
```

```
  ...
)

## S3 method for class 'gam'
n_parameters(
  x,
  component = c("all", "conditional", "smooth_terms"),
  remove_nonestimable = FALSE,
  ...
)

## S3 method for class 'brmsfit'
n_parameters(x, effects = "all", component = "all", ...)
```

Arguments

`x`	A statistical model.
`...`	Arguments passed to or from other methods.
`remove_nonestimable`	Logical, if `TRUE`, removes (i.e. does not count) non-estimable parameters (which may occur for models with rank-deficient model matrix).
`effects`	Should number of parameters for fixed effects, random effects or both be returned? Only applies to mixed models. May be abbreviated.
`component`	Should total number of parameters, number parameters for the conditional model, the zero-inflated part of the model, the dispersion term or the instrumental variables be returned? Applies to models with zero-inflated and/or dispersion formula, or to models with instrumental variable (so called fixed-effects regressions). May be abbreviated.

Value

The number of parameters in the model.

Note

This function returns the number of parameters for the fixed effects by default, as returned by `find_parameters(x, effects = "fixed")`. It does not include *all* estimated model parameters, i.e. auxiliary parameters like sigma or dispersion are not counted. To get the number of *all estimated* parameters, use `get_df(x, type = "model")`.

Examples

```
data(iris)
model <- lm(Sepal.Length ~ Sepal.Width * Species, data = iris)
n_parameters(model)
```

object_has_names *Check names and rownames*

Description

`object_has_names()` checks if specified names are present in the given object. `object_has_rownames()` checks if rownames are present in a dataframe.

Usage

```
object_has_names(x, names)

object_has_rownames(x)
```

Arguments

`x`	A named object (an atomic vector, a list, a dataframe, etc.).
`names`	A single character or a vector of characters.

Value

A logical or a vector of logicals.

Examples

```
# check if specified names are present in the given object
object_has_names(mtcars, "am")
object_has_names(anscombe, c("x1", "z1", "y1"))
object_has_names(list("x" = 1, "y" = 2), c("x", "a"))

# check if a dataframe has rownames
object_has_rownames(mtcars)
```

print_color *Coloured console output*

Description

Convenient function that allows coloured output in the console. Mainly implemented to reduce package dependencies.

Usage

```
print_color(text, color)

print_colour(text, colour)

color_text(text, color)

colour_text(text, colour)

color_theme()
```

Arguments

`text`	The text to print.
`color, colour`	Character vector, indicating the colour for printing. May be one of `"red"`, `"yellow"`, `"green"`, `"blue"`, `"violet"`, `"cyan"` or `"grey"`. Formatting is also possible with `"bold"` or `"italic"`.

Details

This function prints `text` directly to the console using `cat()`, so no string is returned. `color_text()`, however, returns only the formatted string, without using `cat()`. `color_theme()` either returns `"dark"` when RStudio is used with dark color scheme, `"light"` when it's used with light theme, and `NULL` if the theme could not be detected.

Value

Nothing.

Examples

```
print_color("I'm blue dabedi dabedei", "blue")
```

`print_parameters` *Prepare summary statistics of model parameters for printing*

Description

This function takes a data frame, typically a data frame with information on summaries of model parameters like `bayestestR::describe_posterior()`, `bayestestR::hdi()` or `parameters::model_parameters()`, as input and splits this information into several parts, depending on the model. See details below.

Usage

```
print_parameters(
  x,
  ...,
  split_by = c("Effects", "Component", "Group", "Response"),
  format = "text",
  parameter_column = "Parameter",
  keep_parameter_column = TRUE,
  remove_empty_column = FALSE,
  titles = NULL,
  subtitles = NULL
)
```

Arguments

`x` A fitted model, or a data frame returned by `clean_parameters()`.

`...` One or more objects (data frames), which contain information about the model parameters and related statistics (like confidence intervals, HDI, ROPE, ...).

`split_by` `split_by` should be a character vector with one or more of the following elements: `"Effects"`, `"Component"`, `"Response"` and `"Group"`. These are the column names returned by `clean_parameters()`, which is used to extract the information from which the group or component model parameters belong. If `NULL`, the merged data frame is returned. Else, the data frame is split into a list, split by the values from those columns defined in `split_by`.

`format` Name of output-format, as string. If `NULL` (or `"text"`), assumed use for output is basic printing. If `"markdown"`, markdown-format is assumed. This only affects the style of title- and table-caption attributes, which are used in `export_table()`.

`parameter_column` String, name of the column that contains the parameter names. Usually, for data frames returned by functions the easystats-packages, this will be `"Parameter"`.

`keep_parameter_column` Logical, if `TRUE`, the data frames in the returned list have both a `"Cleaned_Parameter"` and `"Parameter"` column. If `FALSE`, the (unformatted) `"Parameter"` is removed, and the column with cleaned parameter names (`"Cleaned_Parameter"`) is renamed into `"Parameter"`.

`remove_empty_column` Logical, if `TRUE`, columns with completely empty character values will be removed.

`titles, subtitles` By default, the names of the model components (like fixed or random effects, count or zero-inflated model part) are added as attributes `"table_title"` and `"table_subtitle"` to each list element returned by `print_parameters()`. These attributes are then extracted and used as table (sub) titles in `export_table()`. Use `titles` and `subtitles` to override the default attribute values for `"table_title"` and `"table_subtitle"`. `titles` and `subtitles` may be any length from 1 to same length as returned list elements. If `titles` and `subtitles` are shorter than existing elements, only the first default attributes are overwritten.

Details

This function prepares data frames that contain information about model parameters for clear printing.

First, `x` is required, which should either be a model object or a prepared data frame as returned by `clean_parameters()`. If `x` is a model, `clean_parameters()` is called on that model object to get information with which model components the parameters are associated.

Then, `...` take one or more data frames that also contain information about parameters from the same model, but also have additional information provided by other methods. For instance, a data frame in `...` might be the result of, for instance, `bayestestR::describe_posterior()`, or `parameters::model_parameters()`, where we have a) a `Parameter` column and b) columns with other parameter values (like CI, HDI, test statistic, etc.).

Now we have a data frame with model parameters and information about the association to the different model components, a data frame with model parameters, and some summary statistics. `print_parameters()` then merges these data frames, so the parameters or statistics of interest are also associated with the different model components. The data frame is split into a list, so for a clear printing. Users can loop over this list and print each component for a better overview. Further, parameter names are "cleaned", if necessary, also for a cleaner print. See also 'Examples'.

Value

A data frame or a list of data frames (if `split_by` is not `NULL`). If a list is returned, the element names reflect the model components where the extracted information in the data frames belong to, e.g. `random.zero_inflated.Intercept: persons`. This is the data frame that contains the parameters for the random effects from group-level "persons" from the zero-inflated model component.

Examples

```
## Not run:
library(bayestestR)
model <- download_model("brms_zi_2")
x <- hdi(model, effects = "all", component = "all")

# hdi() returns a data frame; here we use only the
# information on parameter names and HDI values
tmp <- as.data.frame(x)[, 1:4]
tmp

# Based on the "split_by" argument, we get a list of data frames that
# is split into several parts that reflect the model components.
print_parameters(model, tmp)

# This is the standard print()-method for "bayestestR::hdi"-objects.
# For printing methods, it is easy to print complex summary statistics
# in a clean way to the console by splitting the information into
# different model components.
x
```

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

---

standardize_column_order   *Standardize column order*

Description

Standardizes order of columns for dataframes and other objects from *easystats* and *broom* ecosystem packages.

Usage

```
standardize_column_order(data, ...)

## S3 method for class 'parameters_model'
standardize_column_order(data, style = c("easystats", "broom"), ...)
```

Arguments

data	A data frame. In particular, objects from *easystats* package functions like `parameters::model_parameters()` or `effectsize::effectsize()` are accepted, but also data frames returned by `broom::tidy()` are valid objects.
...	Currently not used.
style	Standardization can either be based on the naming conventions from the easystats-project, or on **broom**'s naming scheme.

Value

A data frame, with standardized column order.

Examples

```
# easystats conventions
df1 <- cbind.data.frame(
  CI_low      = -2.873,
  t           = 5.494,
  CI_high     = -1.088,
  p           = 0.00001,
  Parameter   = -1.980,
  CI          = 0.95,
  df          = 29.234,
  Method      = "Student's t-test"
)

standardize_column_order(df1, style = "easystats")

# broom conventions
```

```
df2 <- cbind.data.frame(
  conf.low   = -2.873,
  statistic  = 5.494,
  conf.high  = -1.088,
  p.value    = 0.00001,
  estimate   = -1.980,
  conf.level = 0.95,
  df         = 29.234,
  method     = "Student's t-test"
)

standardize_column_order(df2, style = "broom")
```

---

standardize_names *Standardize column names*

Description

Standardize column names from data frames, in particular objects returned from `parameters::model_parameters()`, so column names are consistent and the same for any model object.

Usage

```
standardize_names(data, ...)

## S3 method for class 'parameters_model'
standardize_names(
  data,
  style = c("easystats", "broom"),
  ignore_estimate = FALSE,
  ...
)
```

Arguments

`data`	A data frame. In particular, objects from *easystats* package functions like `parameters::model_parameters()` or `effectsize::effectsize()` are accepted, but also data frames returned by `broom::tidy()` are valid objects.
`...`	Currently not used.
`style`	Standardization can either be based on the naming conventions from the easystats-project, or on **broom**'s naming scheme.
`ignore_estimate`	Logical, if `TRUE`, column names like `"mean"` or `"median"` will *not* be converted to `"Coefficient"` resp. `"estimate"`.

Details

This method is in particular useful for package developers or users who use, e.g., parameters::model_parameters() in their own code or functions to retrieve model parameters for further processing. As model_parameters() returns a data frame with varying column names (depending on the input), accessing the required information is probably not quite straightforward. In such cases, standardize_names() can be used to get consistent, i.e. always the same column names, no matter what kind of model was used in model_parameters().

For style = "broom", column names are renamed to match **broom**'s naming scheme, i.e. Parameter is renamed to term, Coefficient becomes estimate and so on.

For style = "easystats", when data is an object from broom::tidy(), column names are converted from "broom"-style into "easystats"-style.

Value

A data frame, with standardized column names.

Examples

```
if (require("parameters")) {
  model <- lm(mpg ~ wt + cyl, data = mtcars)
  mp <- model_parameters(model)

  as.data.frame(mp)
  standardize_names(mp)
  standardize_names(mp, style = "broom")
}
```

---

text_remove_backticks   *Remove backticks from a string*

Description

This function removes backticks from a string.

Usage

```
text_remove_backticks(x, ...)

## S3 method for class 'data.frame'
text_remove_backticks(x, column = "Parameter", verbose = FALSE, ...)
```

Arguments

`x`	A character vector, a data frame or a matrix. If a matrix, backticks are removed from the column and row names, not from values of a character vector.
`...`	Currently not used.
`column`	If `x` is a data frame, specify the column of character vectors, where backticks should be removed. If `NULL`, all character vectors are processed.
`verbose`	Toggle warnings.

Value

`x`, where all backticks are removed.

Note

If `x` is a character vector or data frame, backticks are removed from the elements of that character vector (or character vectors from the data frame.) If `x` is a matrix, the behaviour slightly differs: in this case, backticks are removed from the column and row names. The reason for this behaviour is that this function mainly serves formatting coefficient names. For `vcov()` (a matrix), row and column names equal the coefficient names and therefore are manipulated then.

Examples

```
# example model
data(iris)
iris$`a m` <- iris$Species
iris$`Sepal Width` <- iris$Sepal.Width
model <- lm(`Sepal Width` ~ Petal.Length + `a m`, data = iris)

# remove backticks from string
names(coef(model))
text_remove_backticks(names(coef(model)))

# remove backticks from character variable in a data frame
# column defaults to "Parameter".
d <- data.frame(
  Parameter = names(coef(model)),
  Estimate = unname(coef(model))
)
d
text_remove_backticks(d)
```

`trim_ws` *Small helper functions*

Description

Collection of small helper functions. `trim_ws()` is an efficient function to trim leading and trailing whitespaces from character vectors or strings. `n_unique()` returns the number of unique values in a vector. `safe_deparse()` is comparable to `deparse1()`, i.e. it can safely deparse very long expressions into a single string.

Usage

```
trim_ws(x, ...)

## S3 method for class 'data.frame'
trim_ws(x, character_only = TRUE, ...)

n_unique(x, ...)

## Default S3 method:
n_unique(x, na.rm = TRUE, ...)

safe_deparse(x, ...)
```

Arguments

`x`	A (character) vector, or for some functions may also be a data frame.
`...`	Currently not used.
`character_only`	Logical, if `TRUE` and `x` is a data frame or list, only processes character vectors.
`na.rm`	Logical, if missing values should be removed from the input.

Value

For a vector, `n_unique` always returns an integer value, even if the input is `NULL` (the return value will be `0` then). For data frames or lists, `n_unique()` returns a named numeric vector, with the number of unique values for each element.

Examples

```
trim_ws("  no space!  ")
n_unique(iris$Species)
```

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