# Package 'datawizard’ 

July 21, 2024
Type Package
Title Easy Data Wrangling and Statistical Transformations
Version 0.12.2
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Description A lightweight package to assist in key steps involved in any data analysis workflow: (1) wrangling the raw data to get it in the needed form, (2) applying preprocessing steps and statistical transformations, and (3) compute statistical summaries of data properties and distributions. It is also the data wrangling backend for packages in 'easystats' ecosystem. References: Patil et al. (2022) [doi:10.21105/joss.04684](doi:10.21105/joss.04684).
License MIT + file LICENSE
URL https://easystats.github.io/datawizard/
BugReports https://github.com/easystats/datawizard/issues
Depends R (>= 3.6)
Imports insight ( $>=0.20 .1$ ), stats, utils
Suggests bayestestR, boot, brms, curl, data.table, dplyr (>= 1.1), effectsize, emmeans, gamm4, ggplot2 (>= 3.5.0), gt, haven, httr, knitr, lme4, mediation, modelbased, parameters ( $>=$ 0.21 .7 ), poorman ( $>=0.2 .7$ ), psych, readxl, readr, rio, rmarkdown, rstanarm, see, testthat ( $>=3.2 .1$ ), tibble, tidyr, withr

VignetteBuilder knitr
Encoding UTF-8
Language en-US
RoxygenNote 7.3.2
Config/testthat/edition 3
Config/testthat/parallel true
Config/Needs/website easystats/easystatstemplate
NeedsCompilation no

```
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Repository CRAN
Date/Publication 2024-07-21 02:30:02 UTC
```


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adjust Adjust data for the effect of other variable(s)

## Description

This function can be used to adjust the data for the effect of other variables present in the dataset. It is based on an underlying fitting of regressions models, allowing for quite some flexibility, such as including factors as random effects in mixed models (multilevel partialization), continuous variables as smooth terms in general additive models (non-linear partialization) and/or fitting these models under a Bayesian framework. The values returned by this function are the residuals of the regression models. Note that a regular correlation between two "adjusted" variables is equivalent to the partial correlation between them.

## Usage

adjust (
data,
effect = NULL,
select = is.numeric,
exclude = NULL,
multilevel = FALSE,
additive = FALSE,
bayesian = FALSE,
keep_intercept = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = FALSE
)
data_adjust(
data,
effect $=$ NULL,
select = is.numeric,
exclude $=$ NULL,
multilevel = FALSE,
additive = FALSE,
bayesian = FALSE,
keep_intercept = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = FALSE
)

## Arguments

$$
\text { data } \quad \text { A data frame. }
$$

effect Character vector of column names to be adjusted for (regressed out). If NULL (the default), all variables will be selected.
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or c(1, 3, 5)),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
multilevel If TRUE, the factors are included as random factors. Else, if FALSE (default), they are included as fixed effects in the simple regression model.
additive If TRUE, continuous variables as included as smooth terms in additive models. The goal is to regress-out potential non-linear effects.
bayesian If TRUE, the models are fitted under the Bayesian framework using rstanarm.
keep_intercept If FALSE (default), the intercept of the model is re-added. This avoids the centering around 0 that happens by default when regressing out another variable (see the examples below for a visual representation of this).
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex $=$ TRUE, select must be a character string (or a variable
containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.


## Value

A data frame comparable to data, with adjusted variables.

## Examples

```
adjusted_all <- adjust(attitude)
head(adjusted_all)
adjusted_one <- adjust(attitude, effect = "complaints", select = "rating")
head(adjusted_one)
adjust(attitude, effect = "complaints", select = "rating", bayesian = TRUE)
adjust(attitude, effect = "complaints", select = "rating", additive = TRUE)
attitude$complaints_LMH <- cut(attitude$complaints, 3)
adjust(attitude, effect = "complaints_LMH", select = "rating", multilevel = TRUE)
# Generate data
data <- simulate_correlation(n = 100, r = 0.7)
data$V2 <- (5 * data$V2) + 20 # Add intercept
# Adjust
adjusted <- adjust(data, effect = "V1", select = "V2")
adjusted_icpt <- adjust(data, effect = "V1", select = "V2", keep_intercept = TRUE)
# Visualize
plot(
    data$V1, data$V2,
    pch = 19, col = "blue",
    ylim = c(min(adjusted$V2), max(data$V2)),
    main = "Original (blue), adjusted (green), and adjusted - intercept kept (red) data"
)
abline(lm(V2 ~ V1, data = data), col = "blue")
points(adjusted$V1, adjusted$V2, pch = 19, col = "green")
abline(lm(V2 ~ V1, data = adjusted), col = "green")
points(adjusted_icpt$V1, adjusted_icpt$V2, pch = 19, col = "red")
abline(lm(V2 ~ V1, data = adjusted_icpt), col = "red")
```


## Description

Assign variable and values labels to a variable or variables in a data frame. Labels are stored as attributes ("label" for variable labels and "labels") for value labels.

## Usage

```
assign_labels(x, ...)
## S3 method for class 'numeric'
assign_labels(x, variable = NULL, values = NULL, ...)
## S3 method for class 'data.frame'
assign_labels(
    x,
    select = NULL,
    exclude = NULL,
    values = NULL,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
)
```


## Arguments

x
... Currently not used.
variable The variable label as string.
values The value labels as (named) character vector. If values is not a named vector, the length of labels must be equal to the length of unique values. For a named vector, the left-hand side (LHS) is the value in $x$, the right-hand side (RHS) the associated value label. Non-matching labels are omitted.
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., $\sim$ column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or $-1:-3$ ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a-, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.


## Value

A labelled variable, or a data frame of labelled variables.

## Selection of variables - the select argument

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

## Examples

```
x <- 1:3
# labelling by providing required number of labels
assign_labels(
    x,
    variable = "My x",
    values = c("one", "two", "three")
```

```
)
# labelling using named vectors
data(iris)
out <- assign_labels(
    iris$Species,
    variable = "Labelled Species",
    values = c(`setosa` = "Spec1", `versicolor` = "Spec2", `virginica` = "Spec3")
)
str(out)
# data frame example
out <- assign_labels(
    iris,
    select = "Species",
    variable = "Labelled Species",
    values = c(`setosa` = "Spec1", `versicolor` = "Spec2", `virginica` = "Spec3")
)
str(out$Species)
# Partial labelling
x <- 1:5
assign_labels(
    x,
    variable = "My x",
    values = c(`1` = "lowest", `5` = "highest")
)
```

    categorize
    Recode (or "cut" / "bin") data into groups of values.

## Description

This functions divides the range of variables into intervals and recodes the values inside these intervals according to their related interval. It is basically a wrapper around base R's cut (), providing a simplified and more accessible way to define the interval breaks (cut-off values).

## Usage

```
categorize(x, ...)
## S3 method for class 'numeric'
categorize(
    x,
    split = "median",
    n_groups = NULL,
    range = NULL,
    lowest = 1,
    labels = NULL,
```

```
    verbose = TRUE,
)
## S3 method for class 'data.frame'
categorize(
    x,
    select = NULL,
    exclude = NULL,
    split = "median",
    n_groups = NULL,
    range = NULL,
    lowest = 1,
    labels = NULL,
    append = FALSE,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
    ..
)
```


## Arguments

x
...
split Character vector, indicating at which breaks to split variables, or numeric values with values indicating breaks. If character, may be one of "median", "mean", "quantile", "equal_length", or "equal_range". "median" or "mean" will return dichotomous variables, split at their mean or median, respectively. "quantile" and "equal_length" will split the variable into n_groups groups, where each group refers to an interval of a specific range of values. Thus, the length of each interval will be based on the number of groups. "equal_range" also splits the variable into multiple groups, however, the length of the interval is given, and the number of resulting groups (and hence, the number of breaks) will be determined by how many intervals can be generated, based on the full range of the variable.
n_groups If split is "quantile" or "equal_length", this defines the number of requested groups (i.e. resulting number of levels or values) for the recoded variable(s). "quantile" will define intervals based on the distribution of the variable, while "equal_length" tries to divide the range of the variable into pieces of equal length.
range If split = "equal_range", this defines the range of values that are recoded into a new value.
lowest Minimum value of the recoded variable(s). If NULL (the default), for numeric variables, the minimum of the original input is preserved. For factors, the default minimum is 1 . For split = "equal_range", the default minimum is always 1, unless specified otherwise in lowest.

| labels | Character vector of value labels. If not NULL, categorize() will returns factors <br> instead of numeric variables, with labels used for labelling the factor levels. <br> Can also be "mean" or "median" for a factor with labels as the mean/median of <br> each groups. |
| :--- | :--- |
| verbose | Toggle warnings. <br> select |
|  | Variables that will be included when performing the required tasks. Can be | either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select $=c($ "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
append
ignore_case
Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to $x$, thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in $x$ will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.

Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex $=$ TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

## Value

$x$, recoded into groups. By default $x$ is numeric, unless labels is specified. In this case, a factor is returned, where the factor levels (i.e. recoded groups are labelled accordingly.

## Splits and breaks (cut-off values)

Breaks are in general exclusive, this means that these values indicate the lower bound of the next group or interval to begin. Take a simple example, a numeric variable with values from 1 to 9 . The median would be 5, thus the first interval ranges from 1-4 and is recoded into 1 , while 5-9 would turn into 2 (compare cbind(1:9, categorize(1:9))). The same variable, using split $=$ "quantile" and n_groups = 3 would define breaks at 3.67 and 6.33 (see quantile (1:9, probs $=c(1 / 3,2 / 3))$ ), which means that values from 1 to 3 belong to the first interval and are recoded into 1 (because the next interval starts at 3.67), 4 to 6 into 2 and 7 to 9 into 3 .

## Recoding into groups with equal size or range

split = "equal_length" and split = "equal_range" try to divide the range of $x$ into intervals of similar (or same) length. The difference is that split = "equal_length" will divide the range of
 cut ( $x$, breaks = n_groups) ), while split = "equal_range" will cut $x$ into intervals that all have the length of range, where the first interval by defaults starts at 1 . The lowest (or starting) value of that interval can be defined using the lowest argument.

## Selection of variables - the select argument

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

## See Also

- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()


## Examples

```
set.seed(123)
x <- sample(1:10, size = 50, replace = TRUE)
table(x)
# by default, at median
table(categorize(x))
# into 3 groups, based on distribution (quantiles)
table(categorize(x, split = "quantile", n_groups = 3))
# into 3 groups, user-defined break
table(categorize(x, split = c(3, 5)))
set.seed(123)
x <- sample(1:100, size = 500, replace = TRUE)
# into 5 groups, try to recode into intervals of similar length,
# i.e. the range within groups is the same for all groups
table(categorize(x, split = "equal_length", n_groups = 5))
# into 5 groups, try to return same range within groups
# i.e. 1-20, 21-40, 41-60, etc. Since the range of " }x\mathrm{ " is
# 1-100, and we have a range of 20, this results into 5
# groups, and thus is for this particular case identical
# to the previous result.
table(categorize(x, split = "equal_range", range = 20))
# return factor with value labels instead of numeric value
set.seed(123)
x <- sample(1:10, size = 30, replace = TRUE)
categorize(x, "equal_length", n_groups = 3)
categorize(x, "equal_length", n_groups = 3, labels = c("low", "mid", "high"))
# cut numeric into groups with the mean or median as a label name
x <- sample(1:10, size = 30, replace = TRUE)
categorize(x, "equal_length", n_groups = 3, labels = "mean")
categorize(x, "equal_length", n_groups = 3, labels = "median")
```

center
Centering (Grand-Mean Centering)

## Description

Performs a grand-mean centering of data.

## Usage

```
center(x, ...)
centre(x, ...)
## S3 method for class 'numeric'
center(
        x,
        robust = FALSE,
        weights = NULL,
        reference = NULL,
        center = NULL,
        verbose = TRUE,
)
    ## S3 method for class 'data.frame'
    center(
        x,
        select = NULL,
        exclude = NULL,
        robust = FALSE,
        weights = NULL,
        reference = NULL,
        center = NULL,
        force = FALSE,
        remove_na = c("none", "selected", "all"),
        append = FALSE,
        ignore_case = FALSE,
        verbose = TRUE,
        regex = FALSE,
    )
```


## Arguments

x
...
robust
weights
reference

A (grouped) data frame, a (numeric or character) vector or a factor.
Currently not used.
Logical, if TRUE, centering is done by subtracting the median from the variables. If FALSE, variables are centered by subtracting the mean.

Can be NULL (for no weighting), or:

- For data frames: a numeric vector of weights, or a character of the name of a column in the data.frame that contains the weights.
- For numeric vectors: a numeric vector of weights.

A data frame or variable from which the centrality and deviation will be computed instead of from the input variable. Useful for standardizing a subset or new data according to another data frame.
$\begin{array}{ll}\text { center } & \begin{array}{l}\text { Numeric value, which can be used as alternative to reference to define a refer- } \\ \text { ence centrality. If center is of length 1, it will be recycled to match the length } \\ \text { of selected variables for centering. Else, center must be of same length as the } \\ \text { number of selected variables. Values in center will be matched to selected vari- } \\ \text { ables in the provided order, unless a named vector is given. In this case, names } \\ \text { are matched against the names of the selected variables. }\end{array} \\ \text { verbose } & \begin{array}{l}\text { Toggle warnings and messages. } \\ \text { select }\end{array} \quad \text { Variables that will be included when performing the required tasks. Can be }\end{array}$ either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function $(x)$ mean $(x)>3$ ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
force Logical, if TRUE, forces centering of factors as well. Factors are converted to numerical values, with the lowest level being the value 1 (unless the factor has numeric levels, which are converted to the corresponding numeric value).

How should missing values (NA) be treated: if "none" (default): each column's standardization is done separately, ignoring NAs. Else, rows with NA in the columns selected with select / exclude ("selected") or in all columns ("all") are dropped before standardization, and the resulting data frame does not include these cases.


#### Abstract

append Logical or string. If TRUE, centered variables get new column names (with the suffix "_c") and are appended (column bind) to $x$, thus returning both the original and the centered variables. If FALSE, original variables in $x$ will be overwritten by their centered versions. If a character value, centered variables are appended with new column names (using the defined suffix) to the original data frame. ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names. regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.


## Value

The centered variables.

## Selection of variables - the select argument

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

## Note

Difference between centering and standardizing: Standardized variables are computed by subtracting the mean of the variable and then dividing it by the standard deviation, while centering variables involves only the subtraction.

## See Also

If centering within-clusters (instead of grand-mean centering) is required, see demean(). For standardizing, see standardize(), and makepredictcall.dw_transformer() for use in model formulas.

## Examples

```
data(iris)
# entire data frame or a vector
head(iris$Sepal.Width)
head(center(iris$Sepal.Width))
head(center(iris))
head(center(iris, force = TRUE))
```

```
    # only the selected columns from a data frame
    center(anscombe, select = c("x1", "x3"))
    center(anscombe, exclude = c("x1", "x3"))
    # centering with reference center and scale
d <- data.frame(
    a = c(-2, -1, 0, 1, 2),
    b = c(3, 4, 5, 6, 7)
)
# default centering at mean
center(d)
# centering, using 0 as mean
center(d, center = 0)
# centering, using -5 as mean
center(d, center = -5)
```

    coef_var Compute the coefficient of variation
    
## Description

Compute the coefficient of variation ( CV , ratio of the standard deviation to the mean, $\sigma / \mu$ ) for a set of numeric values.

## Usage

```
coef_var (x, ...)
distribution_coef_var(x, ...)
\#\# S3 method for class 'numeric'
coef_var (
    x ,
    \(m u=\) NULL,
    sigma \(=\) NULL,
    method = c("standard", "unbiased", "median_mad", "qcd"),
    trim = 0,
    remove_na = FALSE,
    \(\mathrm{n}=\mathrm{NULL}\),
    na.rm = FALSE,
)
```


## Arguments

X
.. Further arguments passed to computation functions.
mu A numeric vector of mean values to use to compute the coefficient of variation. If supplied, $x$ is not used to compute the mean.
sigma A numeric vector of standard deviation values to use to compute the coefficient of variation. If supplied, $x$ is not used to compute the SD.
method Method to use to compute the CV. Can be "standard" to compute by dividing the standard deviation by the mean, "unbiased" for the unbiased estimator for normally distributed data, or one of two robust alternatives: "median_mad" to divide the median by the stats: :mad(), or "qcd" (quartile coefficient of dispersion, interquartile range divided by the sum of the quartiles [twice the midhinge]: $\left(Q_{3}-Q_{1}\right) /\left(Q_{3}+Q_{1}\right)$.
trim the fraction (0 to 0.5 ) of values to be trimmed from each end of $x$ before the mean and standard deviation (or other measures) are computed. Values of trim outside the range of ( 0 to 0.5 ) are taken as the nearest endpoint.
remove_na Logical. Should NA values be removed before computing (TRUE) or not (FALSE, default)?
n
If method = "unbiased" and both mu and sigma are provided (not computed from $x$ ), what sample size to use to adjust the computed CV for small-sample bias?
na.rm Deprecated. Please use remove_na instead.

## Details

CV is only applicable of values taken on a ratio scale: values that have a fixed meaningfully defined 0 (which is either the lowest or highest possible value), and that ratios between them are interpretable For example, how many sandwiches have I eaten this week? 0 means "none" and 20 sandwiches is 4 times more than 5 sandwiches. If I were to center the number of sandwiches, it will no longer be on a ratio scale ( 0 is no "none" it is the mean, and the ratio between 4 and -2 is not meaningful). Scaling a ratio scale still results in a ratio scale. So I can re define "how many half sandwiches did I eat this week ( = sandwiches $* 0.5$ ) and 0 would still mean "none", and 20 half-sandwiches is still 4 times more than 5 half-sandwiches.

This means that CV is NOT invariant to shifting, but it is to scaling:

```
sandwiches <- c(0, 4, 15, 0, 0, 5, 2, 7)
coef_var(sandwiches)
#> [1] 1.239094
coef_var(sandwiches / 2) # same
#> [1] 1.239094
coef_var(sandwiches + 4) # different! 0 is no longer meaningful!
#> [1] 0.6290784
```


## Value

The computed coefficient of variation for $x$.

## Examples

```
coef_var(1:10)
coef_var(c(1:10, 100), method = "median_mad")
coef_var(c(1:10, 100), method = "qcd")
coef_var(mu = 10, sigma = 20)
coef_var(mu = 10, sigma = 20, method = "unbiased", n = 30)
```

```
coerce_to_numeric Convert to Numeric (if possible)
```


## Description

Tries to convert vector to numeric if possible (if no warnings or errors). Otherwise, leaves it as is.

## Usage

coerce_to_numeric(x)

## Arguments

x
A vector to be converted.

## Value

Numeric vector (if possible)

## Examples

```
coerce_to_numeric(c("1", "2"))
coerce_to_numeric(c("1", "2", "A"))
```


## Description

Build a deviation contrast matrix, a type of effects contrast matrix.

## Usage

contr.deviation(n, base $=1$, contrasts $=$ TRUE, sparse $=$ FALSE)

## Arguments

n
base an integer specifying which group is considered the baseline group. Ignored if contrasts is FALSE.
contrasts a logical indicating whether contrasts should be computed.
sparse logical indicating if the result should be sparse (of class dgCMatrix), using package Matrix.

## Details

In effects coding, unlike treatment/dummy coding (stats: : contr.treatment()), each contrast sums to 0 . In regressions models, this results in an intercept that represents the (unweighted) average of the group means. In ANOVA settings, this also guarantees that lower order effects represent main effects (and not simple or conditional effects, as is the case when using R's default stats: :contr.treatment()).

Deviation coding (contr. deviation) is a type of effects coding. With deviation coding, the coefficients for factor variables are interpreted as the difference of each factor level from the base level (this is the same interpretation as with treatment/dummy coding). For example, for a factor group with levels " A ", " B ", and " C ", with contr. devation, the intercept represents the overall mean (average of the group means for the 3 groups), and the coefficients groupB and groupC represent the differences between the A group mean and the B and C group means, respectively.

Sum coding (stats: : contr. sum()) is another type of effects coding. With sum coding, the coefficients for factor variables are interpreted as the difference of each factor level from the grand (across-groups) mean. For example, for a factor group with levels "A", "B", and "C", with contr. sum, the intercept represents the overall mean (average of the group means for the 3 groups), and the coefficients group1 and group2 represent the differences the $\mathbf{A}$ and $\mathbf{B}$ group means from the overall mean, respectively.

## See Also

```
stats::contr.sum()
```


## Examples

```
data("mtcars")
mtcars <- data_modify(mtcars, cyl = factor(cyl))
c.treatment <- cbind(Intercept = 1, contrasts(mtcars$cyl))
solve(c.treatment)
#> 4 6 8
#> Intercept 1 0 0 # mean of the 1st level
#> 6 -1 1 0 # 2nd level - 1st level
#> 8 -1 0 1 # 3rd level - 1st level
contrasts(mtcars$cyl) <- contr.sum
c.sum <- cbind(Intercept = 1, contrasts(mtcars$cyl))
solve(c.sum)
#> 4
#> Intercept 0.333 0.333 0.333 # overall mean
#> 0.667-0.333-0.333 # deviation of 1st from overall mean
#> -0.333 0.667-0.333 # deviation of 2nd from overall mean
```

```
contrasts(mtcars$cyl) <- contr.deviation
c.deviation <- cbind(Intercept = 1, contrasts(mtcars$cyl))
solve(c.deviation)
#> 
#> Intercept 0.333 0.333 0.333 # overall mean
#> 6 -1.000 1.000 0.000 # 2nd level - 1st level
#> 8 -1.000 0.000 1.000 # 3rd level - 1st level
## With Interactions ------------------------------------------
mtcars <- data_modify(mtcars, am = C(am, contr = contr.deviation))
mtcars <- data_arrange(mtcars, select = c("cyl", "am"))
mm <- unique(model.matrix(~ cyl * am, data = mtcars))
rownames(mm) <- c(
    "cyl4.am0", "cyl4.am1", "cyl6.am0",
    "cyl6.am1", "cyl8.am0", "cyl8.am1"
)
solve(mm)
#> cyl4.am0 cyl4.am1 cyl6.am0 cyl6.am1 cyl8.am0 cyl8.am1
#> (Intercept) 0.167 0.167 0.167 0.167 0.167 0.167 # overall mean
#> cyl6 -0.500 -0.500 0.500 0.500 0.000 0.000 # cyl MAIN eff: 2nd - 1st
#> cyl8 -0.500 -0.500 0.000 0.000 0.500 0.500 # cyl MAIN eff: 2nd - 1st
#> am1 
#> cyl6:am1 
#> cyl8:am1 
```

convert_na_to Replace missing values in a variable or a data frame.

## Description

Replace missing values in a variable or a data frame.

## Usage

```
    convert_na_to(x, ...)
    ## S3 method for class 'numeric'
    convert_na_to(x, replacement = NULL, verbose = TRUE, ...)
    ## S3 method for class 'character'
    convert_na_to(x, replacement = NULL, verbose = TRUE, ...)
    ## S3 method for class 'data.frame'
    convert_na_to(
        x,
        select = NULL,
        exclude = NULL,
        replacement = NULL,
        replace_num = replacement,
        replace_char = replacement,
        replace_fac = replacement,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE,
    ...
    )
```


## Arguments

x
...
replacement Numeric or character value that will be used to replace NA.
verbose
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., $\sim$ column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function ( $x$ ) mean $(x)>3$ ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
replace_num Value to replace NA when variable is of type numeric.
replace_char Value to replace NA when variable is of type character.
replace_fac Value to replace NA when variable is of type factor.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex $=$ TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.


## Value

$x$, where NA values are replaced by replacement.

## Selection of variables - the select argument

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged.

In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

## Examples

```
# Convert NA to 0 in a numeric vector
convert_na_to(
    c(9, 3, NA, 2, 3, 1, NA, 8),
    replacement = 0
)
# Convert NA to "missing" in a character vector
convert_na_to(
    c("a", NA, "d", "z", NA, "t"),
    replacement = "missing"
)
### For data frames
test_df <- data.frame(
    x = c(1, 2,NA),
    x2 = c(4, 5, NA),
    y = c("a", "b", NA)
)
# Convert all NA to 0 in numeric variables, and all NA to "missing" in
# character variables
convert_na_to(
    test_df,
    replace_num = 0,
    replace_char = "missing"
)
# Convert a specific variable in the data frame
convert_na_to(
    test_df,
    replace_num = 0,
    replace_char = "missing",
    select = "x"
)
# Convert all variables starting with "x"
convert_na_to(
    test_df,
    replace_num = 0,
    replace_char = "missing",
    select = starts_with("x")
)
# Convert NA to 1 in variable 'x2' and to 0 in all other numeric
# variables
convert_na_to(
    test_df,
```

```
    replace_num = 0,
    select = list(x2 = 1)
)
```

convert_to_na Convert non-missing values in a variable into missing values.

## Description

Convert non-missing values in a variable into missing values.

## Usage

```
    convert_to_na(x, ...)
    ## S3 method for class 'numeric'
    convert_to_na(x, na = NULL, verbose = TRUE, ...)
    ## S3 method for class 'factor'
    convert_to_na(x, na = NULL, drop_levels = FALSE, verbose = TRUE, ...)
    ## S3 method for class 'data.frame'
    convert_to_na(
        x,
        select = NULL,
        exclude = NULL,
        na = NULL,
        drop_levels = FALSE,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE,
        ...
    )
```


## Arguments

x
... Not used.
na Numeric, character vector or logical (or a list of numeric, character vectors or logicals) with values that should be converted to NA. Numeric values applied to numeric vectors, character values are used for factors, character vectors or date variables, and logical values for logical vectors.
verbose Toggle warnings.
drop_levels Logical, for factors, when specific levels are replaced by NA, should unused levels be dropped?
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., $\sim$ column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex(""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.


## Value

$x$, where all values in na are converted to NA.

## Examples

$x$ <- sample(1:6, size $=30$, replace $=$ TRUE)

## x

\# values 4 and 5 to NA
convert_to_na(x, na = 4:5)
\# data frames
set.seed(123)
$x$ <- data.frame(
$a=\operatorname{sample}(1: 6$, size $=20$, replace $=$ TRUE $)$,
$b=$ sample(letters[1:6], size $=20$, replace $=$ TRUE),
c $=$ sample(c(30:33, 99), size $=20$, replace $=$ TRUE $)$
)
\# for all numerics, convert 5 to NA. Character/factor will be ignored.
convert_to_na(x, na = 5)
\# for numerics, 5 to NA, for character/factor, "f" to NA
convert_to_na(x, na = list(6, "f"))
\# select specific variables
convert_to_na( $x$, select $=c(" a ", \quad " b ")$, na $=\operatorname{list}(6, \quad " f "))$
data_addprefix Rename columns and variable names

## Description

Safe and intuitive functions to rename variables or rows in data frames. data_rename() will rename column names, i.e. it facilitates renaming variables data_addprefix() or data_addsuffix() add prefixes or suffixes to column names. data_rename_rows() is a convenient shortcut to add or rename row names of a data frame, but unlike row. names(), its input and output is a data frame, thus, integrating smoothly into a possible pipe-workflow.

## Usage

data_addprefix( data,
pattern,
select = NULL,
exclude $=$ NULL,
ignore_case = FALSE,
regex = FALSE,
verbose $=$ TRUE,
...
)
data_addsuffix(
data,
pattern,
select $=$ NULL,
exclude $=$ NULL,

```
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
)
data_rename(
    data,
    pattern = NULL,
    replacement = NULL,
    safe = TRUE,
    verbose = TRUE,
)
data_rename_rows(data, rows = NULL)
```


## Arguments

data A data frame, or an object that can be coerced to a data frame.
pattern Character vector. For data_rename(), indicates columns that should be selected for renaming. Can be NULL (in which case all columns are selected). For data_addprefix() or data_addsuffix(), a character string, which will be added as prefix or suffix to the column names.
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or $-1:-3$ ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean $(x)>3$ ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to
exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".

| exclude | See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns. |
| :---: | :---: |
| ignore_case | Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names. |
| regex | Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex(""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround. |
| verbose | Toggle warnings and messages. |
|  | Other arguments passed to or from other functions. |
| replacement | Character vector. Indicates the new name of the columns selected in pattern. Can be NULL (in which case column are numbered in sequential order). If not NULL, pattern and replacement must be of the same length. |
| safe | Do not throw error if for instance the variable to be renamed/removed doesn't exist. |
| rows | Vector of row names. |

## Value

A modified data frame.

## See Also

- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()


## Examples

```
    # Add prefix / suffix to all columns
    head(data_addprefix(iris, "NEW_"))
    head(data_addsuffix(iris, "_OLD"))
    # Rename columns
    head(data_rename(iris, "Sepal.Length", "length"))
    # data_rename(iris, "FakeCol", "length", safe=FALSE) # This fails
    head(data_rename(iris, "FakeCol", "length")) # This doesn't
    head(data_rename(iris, c("Sepal.Length", "Sepal.Width"), c("length", "width")))
    # Reset names
    head(data_rename(iris, NULL))
    # Change all
    head(data_rename(iris, replacement = paste0("Var", 1:5)))
```

    data_arrange Arrange rows by column values
    
## Description

data_arrange() orders the rows of a data frame by the values of selected columns.

## Usage

data_arrange(data, select $=$ NULL, safe $=$ TRUE)

## Arguments

$$
\begin{array}{ll}
\text { data } & \text { A data frame, or an object that can be coerced to a data frame. } \\
\text { select } & \begin{array}{l}
\text { Character vector of column names. Use a dash just before column name to } \\
\text { arrange in decreasing order, for example " }-\times 1 \text { ". }
\end{array} \\
\text { safe } & \text { Do not throw an error if one of the variables specified doesn't exist. }
\end{array}
$$

## Value

A data frame.

## Examples

```
# Arrange using several variables
data_arrange(head(mtcars), c("gear", "carb"))
# Arrange in decreasing order
data_arrange(head(mtcars), "-carb")
# Throw an error if one of the variables specified doesn't exist
try(data_arrange(head(mtcars), c("gear", "foo"), safe = FALSE))
```

data_codebook Generate a codebook of a data frame.

## Description

data_codebook () generates codebooks from data frames, i.e. overviews of all variables and some more information about each variable (like labels, values or value range, frequencies, amount of missing values).

## Usage

data_codebook(
data,
select $=$ NULL,
exclude = NULL,
variable_label_width = NULL,
value_label_width = NULL,
max_values = 10,
range_at $=6$,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)
\#\# S3 method for class 'data_codebook'
print_html(
x,
font_size = "100\%",
line_padding = 3,
row_color = "\#eeeeee",
)

## Arguments

data A data frame, or an object that can be coerced to a data frame.
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or $-1:-3$ ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select $=c($ "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.

```
variable_label_width
```

Length of variable labels. Longer labels will be wrapped at variable_label_width chars. If NULL, longer labels will not be split into multiple lines. Only applies to labelled data.

Length of value labels. Longer labels will be shortened, where the remaining part is truncated. Only applies to labelled data or factor levels.
max_values Number of maximum values that should be displayed. Can be used to avoid too many rows when variables have lots of unique values.
range_at Indicates how many unique values in a numeric vector are needed in order to print a range for that variable instead of a frequency table for all numeric values. Can be useful if the data contains numeric variables with only a few unique values and where full frequency tables instead of value ranges should be displayed.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings and messages on or off.
... Arguments passed to or from other methods.
$x \quad$ A (grouped) data frame, a vector or a statistical model (for unstandardize() cannot be a model).
font_size For HTML tables, the font size.
line_padding For HTML tables, the distance (in pixel) between lines.
row_color For HTML tables, the fill color for odd rows.

## Value

A formatted data frame, summarizing the content of the data frame. Returned columns include the column index of the variables in the original data frame (ID), column name, variable label (if data is labelled), type of variable, number of missing values, unique values (or value range), value labels (for labelled data), and a frequency table ( N for each value). Most columns are formatted as character vectors.

## Note

There are methods to print() the data frame in a nicer output, as well methods for printing in markdown or HTML format (print_md() and print_html()).

## Examples

```
data(iris)
data_codebook(iris, select = starts_with("Sepal"))
data(efc)
data_codebook(efc)
# shorten labels
data_codebook(efc, variable_label_width = 20, value_label_width = 15)
# automatic range for numerics at more than 5 unique values
data(mtcars)
data_codebook(mtcars, select = starts_with("c"))
# force all values to be displayed
data_codebook(mtcars, select = starts_with("c"), range_at = 100)
```

data_duplicated Extract all duplicates

## Description

Extract all duplicates, for visual inspection. Note that it also contains the first occurrence of future duplicates, unlike duplicated() or dplyr::distinct()). Also contains an additional column reporting the number of missing values for that row, to help in the decision-making when selecting which duplicates to keep.

## Usage

```
data_duplicated(
        data,
        select = NULL,
        exclude = NULL,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE
    )
```


## Arguments

data
select
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex(""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.

## Value

A dataframe, containing all duplicates.

## See Also

```
data_unique()
```


## Examples

```
df1 <- data.frame(
    id = c(1, 2, 3, 1, 3),
    year = c(2022, 2022, 2022, 2022, 2000),
    item1 = c(NA, 1, 1, 2, 3),
    item2 = c(NA, 1, 1, 2, 3),
    item3 = c(NA, 1, 1, 2, 3)
)
data_duplicated(df1, select = "id")
data_duplicated(df1, select = c("id", "year"))
# Filter to exclude duplicates
df2 <- df1[-c(1, 5), ]
df2
```

data_extract

## Description

data_extract() (or its alias extract()) is similar to \$. It extracts either a single column or element from an object (e.g., a data frame, list), or multiple columns resp. elements.

## Usage

```
data_extract(data, select, ...)
    ## S3 method for class 'data.frame'
    data_extract(
        data,
        select,
        name = NULL,
        extract = "all",
        as_data_frame = FALSE,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE,
)
```


## Arguments

data The object to subset. Methods are currently available for data frames and data frame extensions (e.g., tibbles).
select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., $\sim$ column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or $-1:-3$ ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean $(x)>3$ ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".

| $\ldots$ | For use by future methods. |
| :--- | :--- |
| name | An optional argument that specifies the column to be used as names for the <br> vector elements after extraction. Must be specified either as literal variable name <br> (e.g., column_name) or as string ("column_name"). name will be ignored when |
| a data frame is returned. |  |

## Details

data_extract () can be used to select multiple variables or pull a single variable from a data frame. Thus, the return value is by default not type safe - data_extract() either returns a vector or a data frame.

Extracting single variables (vectors): When select is the name of a single column, or when select only matches one column, a vector is returned. A single variable is also returned when extract is either "first or "last". Setting as_data_frame to TRUE overrides this behaviour and always returns a data frame.

Extracting a data frame of variables: When select is a character vector containing more than one column name (or a numeric vector with more than one valid column indices), or when select uses one of the supported select-helpers that match multiple columns, a data frame is returned. Setting as_data_frame to TRUE always returns a data frame.

## Value

A vector (or a data frame) containing the extracted element, or NULL if no matching variable was found.

## Examples

```
# single variable
data_extract(mtcars, cyl, name = gear)
data_extract(mtcars, "cyl", name = gear)
data_extract(mtcars, -1, name = gear)
data_extract(mtcars, cyl, name = 0)
data_extract(mtcars, cyl, name = "row.names")
# selecting multiple variables
head(data_extract(iris, starts_with("Sepal")))
head(data_extract(iris, ends_with("Width")))
head(data_extract(iris, 2:4))
# select first of multiple variables
data_extract(iris, starts_with("Sepal"), extract = "first")
# select first of multiple variables, return as data frame
head(data_extract(iris, starts_with("Sepal"), extract = "first", as_data_frame = TRUE))
```

data_group $\quad$ Create a grouped data frame

## Description

This function is comparable to dplyr::group_by(), but just following the datawizard function design. data_ungroup() removes the grouping information from a grouped data frame.

## Usage

```
data_group(
    data,
    select = NULL,
    exclude = NULL,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
    ...
)
    data_ungroup(data, verbose = TRUE, ...)
```


## Arguments

## data

select Variables that will be included when performing the required tasks. Can be either

- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or $c(1,3,5)$ ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or $-1:-3$ ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean $(x)>3$ ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.
... Arguments passed down to other functions. Mostly not used yet.


## Value

A grouped data frame, i.e. a data frame with additional information about the grouping structure saved as attributes.

## Examples

```
data(efc)
suppressPackageStartupMessages(library(poorman, quietly = TRUE))
# total mean
efc %>%
        summarize(mean_hours = mean(c12hour, na.rm = TRUE))
# mean by educational level
efc %>%
    data_group(c172code) %>%
    summarize(mean_hours = mean(c12hour, na.rm = TRUE))
```

data_match Return filtered or sliced data frame, or row indices

## Description

Return a filtered (or sliced) data frame or row indices of a data frame that match a specific condition. data_filter() works like data_match(), but works with logical expressions or row indices of a data frame to specify matching conditions.

## Usage

data_match(x, to, match = "and", return_indices = FALSE, drop_na = TRUE, ...)
data_filter(x, ...)

## Arguments

$x \quad$ A data frame.
to A data frame matching the specified conditions. Note that if match is a value other than "and", the original row order might be changed. See 'Details'.
match String, indicating with which logical operation matching conditions should be combined. Can be "and" (or "\&"), "or" (or "|") or "not" (or "!").
return_indices Logical, if FALSE, return the vector of rows that can be used to filter the original data frame. If FALSE (default), returns directly the filtered data frame instead of the row indices.
drop_na Logical, if TRUE, missing values (NAs) are removed before filtering the data. This is the default behaviour, however, sometimes when row indices are requested (i.e. return_indices=TRUE), it might be useful to preserve NA values, so returned row indices match the row indices of the original data frame.
... A sequence of logical expressions indicating which rows to keep, or a numeric vector indicating the row indices of rows to keep. Can also be a string representation of a logical expression (e.g. " $x>4$ "), a character vector (e.g. c("x $>4 ", " y==2 ")$ ) or a variable that contains the string representation of a logical expression. These might be useful when used in packages to avoid defining undefined global variables.

## Details

For data_match(), if match is either "or" or "not", the original row order from $\times$ might be changed. If preserving row order is required, use data_filter() instead.

```
# mimics subset() behaviour, preserving original row order
head(data_filter(mtcars[c("mpg", "vs", "am")], vs == 0 | am == 1))
#> mpg vs am
#> Mazda RX4 21.0 0 1
#> Mazda RX4 Wag 21.0 0 1
#> Datsun 710 22.8 1 1
#> Hornet Sportabout 18.7 0 0
#> Duster 360 14.3 0 0
#> Merc 450SE 16.4 0 0
# re-sorting rows
head(data_match(mtcars[c("mpg", "vs", "am")],
    data.frame(vs = 0, am = 1),
        match = "or"))
#> mpg vs am
#> Mazda RX4 21.0 0 1
#> Mazda RX4 Wag 21.0 0 1
#> Hornet Sportabout 18.7 0 0
#> Duster 360 14.3 0 0
#> Merc 450SE 16.4 0 0
#> Merc 450SL 17.3 0 0
```

While data_match() works with data frames to match conditions against, data_filter() is basically a wrapper around subset (subset = <filter>). However, unlike subset(), it preserves label attributes and is useful when working with labelled data.

## Value

A filtered data frame, or the row indices that match the specified configuration.

## See Also

- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder (), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()


## Examples

```
data_match(mtcars, data.frame(vs = 0, am = 1))
data_match(mtcars, data.frame(vs = 0, am = c(0, 1)))
# observations where "vs" is NOT 0 AND "am" is NOT 1
data_match(mtcars, data.frame(vs = 0, am = 1), match = "not")
# equivalent to
data_filter(mtcars, vs != 0 & am != 1)
# observations where EITHER "vs" is 0 OR "am" is 1
data_match(mtcars, data.frame(vs = 0, am = 1), match = "or")
# equivalent to
data_filter(mtcars, vs == 0 | am == 1)
# slice data frame by row indices
data_filter(mtcars, 5:10)
# Define a custom function containing data_filter()
my_filter <- function(data, variable) {
    data_filter(data, variable)
}
my_filter(mtcars, "cyl == 6")
# Pass complete filter-condition as string.
my_filter <- function(data, condition) {
    data_filter(data, condition)
}
my_filter(mtcars, "am != 0")
# string can also be used directly as argument
data_filter(mtcars, "am != 0")
# or as variable
fl <- "am != 0"
data_filter(mtcars, fl)
```

data_merge Merge (join) two data frames, or a list of data frames

## Description

Merge (join) two data frames, or a list of data frames. However, unlike base R's merge(), data_merge() offers a few more methods to join data frames, and it does not drop data frame nor column attributes.

## Usage

```
data_merge(x, ...)
data_join(x, ...)
\#\# S3 method for class 'data.frame'
data_merge(x, y, join = "left", by = NULL, id = NULL, verbose = TRUE, ...)
\#\# S3 method for class 'list'
data_merge(x, join = "left", by = NULL, id = NULL, verbose = TRUE, ...)
```


## Arguments

$x, y \quad$ A data frame to merge. $x$ may also be a list of data frames that will be merged. Note that the list-method has no $y$ argument.
... Not used.
join Character vector, indicating the method of joining the data frames. Can be "full", "left" (default), "right", "inner", "anti", "semi" or "bind". See details below.
by Specifications of the columns used for merging.
id Optional name for ID column that will be created to indicate the source data frames for appended rows. Only applies if join = "bind".
verbose Toggle warnings.

## Value

A merged data frame.

## Merging data frames

Merging data frames is performed by adding rows (cases), columns (variables) or both from the source data frame ( $y$ ) to the target data frame ( $x$ ). This usually requires one or more variables which are included in both data frames and that are used for merging, typically indicated with the by argument. When by contains a variable present in both data frames, cases are matched and filtered by identical values of by in $x$ and $y$.

## Left- and right-joins

Left- and right joins usually don't add new rows (cases), but only new columns (variables) for existing cases in x . For join = "left" or join = "right" to work, by must indicate one or more columns that are included in both data frames. For join = "left", if by is an identifier variable, which is included in both $x$ and $y$, all variables from $y$ are copied to $x$, but only those cases from $y$ that have matching values in their identifier variable in $x$ (i.e. all cases in $x$ that are also found in $y$ get the related values from the new columns in $y$ ). If there is no match between identifiers in $x$ and $y$, the copied variable from $y$ will get a NA value for this particular case. Other variables that occur both in $x$ and $y$, but are not used as identifiers (with by), will be renamed to avoid multiple identical variable names. Cases in $y$ where values from the identifier have no match in $x$ 's identifier
are removed. join = "right" works in a similar way as join = "left", just that only cases from $x$ that have matching values in their identifier variable in $y$ are chosen.
In base $R$, these are equivalent to merge $(x, y$, all. $x=\operatorname{TRUE})$ and merge $(x, y, a l l . y=$ TRUE $)$.

## Full joins

Full joins copy all cases from $y$ to $x$. For matching cases in both data frames, values for new variables are copied from $y$ to $x$. For cases in $y$ not present in $x$, these will be added as new rows to $x$. Thus, full joins not only add new columns (variables), but also might add new rows (cases).
In base $R$, this is equivalent to merge $(x, y$, all $=$ TRUE $)$.

## Inner joins

Inner joins merge two data frames, however, only those rows (cases) are kept that are present in both data frames. Thus, inner joins usually add new columns (variables), but also remove rows (cases) that only occur in one data frame.
In base $R$, this is equivalent to merge $(x, y)$.

## Binds

join = "bind" row-binds the complete second data frame $y$ to $x$. Unlike simple rbind(), which requires the same columns for both data frames, join = "bind" will bind shared columns from y to $x$, and add new columns from $y$ to $x$.

## See Also

- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()


## Examples

```
x <- data.frame(a = 1:3, b = c("a", "b", "c"), c = 5:7, id = 1:3)
y <- data.frame(c = 6:8, d = c("f", "g", "h"), e = 100:102, id = 2:4)
x
y
# "by" will default to all shared columns, i.e. "c" and "id". new columns
# "d" and "e" will be copied from "y" to "x", but there are only two cases
# in "x" that have the same values for "c" and "id" in "y". only those cases
```

```
# have values in the copied columns, the other case gets "NA".
data_merge(x, y, join = "left")
# we change the id-value here
x <- data.frame(a = 1:3, b = c("a", "b", "c"), c = 5:7, id = 1:3)
y <- data.frame(c = 6:8, d = c("f", "g", "h"), e = 100:102, id = 3:5)
x
y
# no cases in "y" have the same matching "c" and "id" as in "x", thus
# copied variables from "y" to "x" copy no values, all get NA.
data_merge(x, y, join = "left")
# one case in "y" has a match in "id" with "x", thus values for this
# case from the remaining variables in "y" are copied to "x", all other
# values (cases) in those remaining variables get NA
data_merge(x, y, join = "left", by = "id")
data(mtcars)
x <- mtcars[1:5, 1:3]
y <- mtcars[28:32, 4:6]
# add ID common column
x$id <- 1:5
y$id <- 3:7
# left-join, add new variables and copy values from y to x,
# where "id" values match
data_merge(x, y)
# right-join, add new variables and copy values from x to y,
# where "id" values match
data_merge(x, y, join = "right")
# full-join
data_merge(x, y, join = "full")
data(mtcars)
x <- mtcars[1:5, 1:3]
y <- mtcars[28:32, c(1, 4:5)]
# add ID common column
x$id <- 1:5
y$id <- 3:7
# left-join, no matching rows (because columns "id" and "disp" are used)
# new variables get all NA values
data_merge(x, y)
# one common value in "mpg", so one row from y is copied to x
data_merge(x, y, by = "mpg")
```

```
# only keep rows with matching values in by-column
data_merge(x, y, join = "semi", by = "mpg")
# only keep rows with non-matching values in by-column
data_merge(x, y, join = "anti", by = "mpg")
# merge list of data frames. can be of different rows
x <- mtcars[1:5, 1:3]
y <- mtcars[28:31, 3:5]
z <- mtcars[11:18, c(1, 3:4, 6:8)]
x$id <- 1:5
y$id <- 4:7
z$id <- 3:10
data_merge(list(x, y, z), join = "bind", by = "id", id = "source")
```

```
data_modify

\section*{Description}

Create new variables or modify existing variables in a data frame. Unlike base: :transform(), data_modify () can be used on grouped data frames, and newly created variables can be directly used.

\section*{Usage}
data_modify(data, ...)
\#\# S3 method for class 'data.frame'
data_modify(data, ..., .if = NULL, .at \(=\) NULL,. modify \(=\) NULL)

\section*{Arguments}

\section*{data}

A data frame
One or more expressions that define the new variable name and the values or recoding of those new variables. These expressions can be one of:
- A sequence of named, literal expressions, where the left-hand side refers to the name of the new variable, while the right-hand side represent the values of the new variable. Example: Sepal. Width = center (Sepal. Width).
- A sequence of string values, representing expressions.
- A variable that contains a string representation of the expression. Example:
```

a <- "2 * Sepal.Width"
data_modify(iris, a)

```
- A character vector of expressions. Example: c("SW_double \(=2 *\) Sepal. Width", "SW_fraction = SW_double / 10"). This type of expression cannot be mixed with other expressions, i.e. if a character vector is provided, you may not add further elements to ....
- Using NULL as right-hand side removes a variable from the data frame. Example: Petal. Width = NULL.
Note that newly created variables can be used in subsequent expressions, including . at or .if. See also 'Examples'.
.if A function that returns TRUE for columns in the data frame where .if applies. This argument is used in combination with the .modify argument. Note that only one of . at or .if can be provided, but not both at the same time. Newly created variables in . . . can also be selected, see 'Examples'.
.at A character vector of variable names that should be modified. This argument is used in combination with the .modify argument. Note that only one of . at or .if can be provided, but not both at the same time. Newly created variables in . . . can also be selected, see 'Examples'.
.modify A function that modifies the variables defined in .at or .if. This argument is used in combination with either the .at or the .if argument. Note that the modified variable (i.e. the result from .modify) must be either of length 1 or of same length as the input variable.

\section*{Note}
data_modify () can also be used inside functions. However, it is recommended to pass the recodeexpression as character vector or list of characters.

\section*{Examples}
```

data(efc)
new_efc <- data_modify(
efc,
c12hour_c = center(c12hour),
c12hour_z = c12hour_c / sd(c12hour, na.rm = TRUE),
c12hour_z2 = standardize(c12hour)
)
head(new_efc)

# using strings instead of literal expressions

new_efc <- data_modify(
efc,
"c12hour_c = center(c12hour)",
"c12hour_z = c12hour_c / sd(c12hour, na.rm = TRUE)",
"c12hour_z2 = standardize(c12hour)"
)
head(new_efc)

# using character strings, provided as variable

stand <- "c12hour_c / sd(c12hour, na.rm = TRUE)"
new_efc <- data_modify(
efc,
c12hour_c = center(c12hour),
c12hour_z = stand
)
head(new_efc)

```
```


# providing expressions as character vector

new_exp <- c(
"c12hour_c = center(c12hour)",
"c12hour_z = c12hour_c / sd(c12hour, na.rm = TRUE)"
)
new_efc <- data_modify(efc, new_exp)
head(new_efc)

# attributes - in this case, value and variable labels - are preserved

str(new_efc)

# overwrite existing variable, remove old variable

out <- data_modify(iris, Petal.Length = 1 / Sepal.Length, Sepal.Length = NULL)
head(out)

# works on grouped data

grouped_efc <- data_group(efc, "c172code")
new_efc <- data_modify(
grouped_efc,
c12hour_c = center(c12hour),
c12hour_z = c12hour_c / sd(c12hour, na.rm = TRUE),
c12hour_z2 = standardize(c12hour)
)
head(new_efc)

# works from inside functions

foo <- function(data, z) {
head(data_modify(data, z))
}
foo(iris, "var_a = Sepal.Width / 10")
new_exp <- c("SW_double = 2 * Sepal.Width", "SW_fraction = SW_double / 10")
foo(iris, new_exp)

# modify at specific positions or if condition is met

d <- iris[1:5, ]
data_modify(d, .at = "Species", .modify = as.numeric)
data_modify(d, .if = is.factor, .modify = as.numeric)

# can be combined with dots

data_modify(d, new_length = Petal.Length * 2, .at = "Species", .modify = as.numeric)

# new variables used in `.at` or `.if`

data_modify(
d,
new_length = Petal.Length * 2,
.at = c("Petal.Length", "new_length"),
.modify = round
)

# combine "extract_column_names()" and ".at" argument

out <- data_modify(

```
d,
.at = extract_column_names(d, select = starts_with("Sepal")), .modify = as.factor
)
\# "Sepal.Length" and "Sepal.Width" are now factors
str (out)
data_partition Partition data

\section*{Description}

Creates data partitions (for instance, a training and a test set) based on a data frame that can also be stratified (i.e., evenly spread a given factor) using the by argument.
```

Usage
data_partition(
data,
proportion = 0.7,
by = NULL,
seed = NULL,
row_id = ".row_id",
verbose = TRUE,
group = NULL,
)

```

\section*{Arguments}
data A data frame, or an object that can be coerced to a data frame.
proportion
by A character vector indicating the name(s) of the column(s) used for stratified partitioning.
seed A random number generator seed. Enter an integer (e.g. 123) so that the random sampling will be the same each time you run the function.
row_id Character string, indicating the name of the column that contains the row-id's.
verbose Toggle messages and warnings.
group Deprecated. Use by instead.
\(\ldots \quad\) Other arguments passed to or from other functions.

\section*{Value}

A list of data frames. The list includes one training set per given proportion and the remaining data as test set. List elements of training sets are named after the given proportions (e.g., \$p_0.7), the test set is named \$test.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```

data(iris)
out <- data_partition(iris, proportion = 0.9)
out$test
nrow(out$p_0.9)

# Stratify by group (equal proportions of each species)

out <- data_partition(iris, proportion = 0.9, by = "Species")
out\$test

# Create multiple partitions

out <- data_partition(iris, proportion = c(0.3, 0.3))
lapply(out, head)

# Create multiple partitions, stratified by group - 30% equally sampled

# from species in first training set, 50% in second training set and

# remaining 20% equally sampled from each species in test set.

out <- data_partition(iris, proportion = c(0.3, 0.5), by = "Species")
lapply(out, function(i) table(i\$Species))

```
data_peek

Peek at values and type of variables in a data frame

\section*{Description}

This function creates a table a data frame, showing all column names, variable types and the first values (as many as fit into the screen).

\section*{Usage}
```

data_peek(x, ...)
\#\# S3 method for class 'data.frame'
data_peek(
x ,
select $=$ NULL,
exclude $=$ NULL,
ignore_case = FALSE,
regex = FALSE,
width = NULL,
verbose = TRUE,
)

```

\section*{Arguments}
x
...
select

A data frame.
not used.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".
\begin{tabular}{ll} 
exclude & \begin{tabular}{l} 
See select, however, column names matched by the pattern from exclude will \\
be excluded instead of selected. If NULL (the default), excludes no columns.
\end{tabular} \\
ignore_case & \begin{tabular}{l} 
Logical, if TRUE and when one of the select-helpers or a regular expression is \\
used in select, ignores lower/upper case in the search pattern when matching \\
against variable names.
\end{tabular} \\
regex & \begin{tabular}{l} 
Logical, if TRUE, the search pattern from select will be treated as regular ex- \\
pression. When regex = TRUE, select must be a character string (or a variable \\
containing a character string) and is not allowed to be one of the supported \\
select-helpers or a character vector of length > 1. regex = TRUE is compara- \\
ble to using one of the two select-helpers, select = contains("") or select \\
= regex(""), however, since the select-helpers may not work when called from \\
inside other functions (see 'Details'), this argument may be used as workaround.
\end{tabular} \\
width & \begin{tabular}{l} 
Maximum width of line length to display. If NULL, width will be determined \\
using options() \$width.
\end{tabular} \\
verbose & \begin{tabular}{l} 
Toggle warnings.
\end{tabular}
\end{tabular}

\section*{Value}

A data frame with three columns, containing information about the name, type and first values of the input data frame.

\section*{Note}

To show only specific or a limited number of variables, use the select argument, e.g. select \(=\) \(1: 5\) to show only the first five variables.

\section*{Examples}
```

data(efc)
data_peek(efc)

# show variables two to four

data_peek(efc, select = 2:4)

```

\section*{Description}

This functions imports data from various file types. It is a small wrapper around haven: : read_spss(), haven::read_stata(), haven::read_sas(), readxl::read_excel() and data.table::fread() resp. readr: : read_delim() (the latter if package data.table is not installed). Thus, supported file types for importing data are data files from SPSS, SAS or Stata, Excel files or text files (like '.csv’ files). All other file types are passed to rio: :import(). data_write() works in a similar way.
```

Usage
data_read(
path,
path_catalog = NULL,
encoding = NULL,
convert_factors = TRUE,
verbose = TRUE,
...
)
data_write(
data,
path,
delimiter = ",",
convert_factors = FALSE,
save_labels = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}
path \(\quad\) Character string, the file path to the data file.
path_catalog Character string, path to the catalog file. Only relevant for SAS data files.
encoding The character encoding used for the file. Usually not needed.
convert_factors
If TRUE (default), numeric variables, where all values have a value label, are assumed to be categorical and converted into factors. If FALSE, no variable types are guessed and no conversion of numeric variables into factors will be performed. See also section 'Differences to other packages'. For data_write(), this argument only applies to the text (e.g. .txt or .csv) or spreadsheet file formats (like .xlsx). Converting to factors might be useful for these formats because labelled numeric variables are then converted into factors and exported as character columns - else, value labels would be lost and only numeric values are written to the file.
verbose Toggle warnings and messages.
... Arguments passed to the related read_*() or write_*() functions.
data The data frame that should be written to a file.
delimiter For CSV-files, specifies the delimiter. Defaults to ", ", but in particular in European regions, "; " might be a useful alternative, especially when exported CSVfiles should be opened in Excel.
save_labels Only applies to CSV files. If TRUE, value and variable labels (if any) will be saved as additional CSV file. This file has the same file name as the exported CSV file, but includes a "_labels" suffix (i.e. when the file name is "mydat.csv", the additional file with value and variable labels is named "mydat_labels.csv").

Value
A data frame.

\section*{Supported file types}
- data_read() is a wrapper around the haven, data.table, readr readxl and rio packages. Currently supported file types are .txt, .csv, .xls, .xlsx, .sav, .por, .dta and .sas (and related files). All other file types are passed to rio: :import().
- data_write() is a wrapper around haven, readr and rio packages, and supports writing files into all formats supported by these packages.

\section*{Compressed files (zip) and URLs}
data_read() can also read the above mentioned files from URLs or from inside zip-compressed files. Thus, path can also be a URL to a file like "http://www.url.com/file.csv". When path points to a zip-compressed file, and there are multiple files inside the zip-archive, then the first supported file is extracted and loaded.

\section*{General behaviour}
data_read() detects the appropriate read_*() function based on the file-extension of the data file. Thus, in most cases it should be enough to only specify the path argument. However, if more control is needed, all arguments in . . . are passed down to the related read_*() function. The same applies to data_write(), i.e. based on the file extension provided in path, the appropriate write_*() function is used automatically.

\section*{SPSS specific behaviour}
data_read() does not import user-defined ("tagged") NA values from SPSS, i.e. argument user_na is always set to FALSE when importing SPSS data with the haven package. Use convert_to_na() to define missing values in the imported data, if necessary. Furthermore, data_write() compresses SPSS files by default. If this causes problems with (older) SPSS versions, use compress = "none", for example data_write(data, "myfile.sav", compress = "none").

\section*{Differences to other packages that read foreign data formats}
data_read() is most comparable to rio: :import(). For data files from SPSS, SAS or Stata, which support labelled data, variables are converted into their most appropriate type. The major difference to rio: :import() is that data_read() automatically converts fully labelled numeric variables into factors, where imported value labels will be set as factor levels. If a numeric variable has no value labels or less value labels than values, it is not converted to factor. In this case, value labels are preserved as "labels" attribute. Character vectors are preserved. Use convert_factors \(=\) FALSE to remove the automatic conversion of numeric variables to factors.
data_relocate Relocate (reorder) columns of a data frame

\section*{Description}
data_relocate() will reorder columns to specific positions, indicated by before or after. data_reorder() will instead move selected columns to the beginning of a data frame. Finally, data_remove() removes columns from a data frame. All functions support select-helpers that allow flexible specification of a search pattern to find matching columns, which should be reordered or removed.

\section*{Usage}
data_relocate( data, select, before = NULL, after = NULL, ignore_case = FALSE, regex = FALSE, verbose = TRUE,
)
data_reorder ( data, select, exclude = NULL, ignore_case = FALSE, regex = FALSE, verbose = TRUE,
)
data_remove( data, select \(=\) NULL, exclude = NULL, ignore_case = FALSE, regex = FALSE, verbose = FALSE,
)

\section*{Arguments}
\[
\text { data } \quad \text { A data frame. }
\]
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function ( \(x\) ) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
before, after Destination of columns. Supplying neither will move columns to the left-hand side; specifying both is an error. Can be a character vector, indicating the name of the destination column, or a numeric value, indicating the index number of the destination column. If -1 , will be added before or after the last column.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.
Arguments passed down to other functions. Mostly not used yet.
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.

\section*{Value}

A data frame with reordered columns.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```


# Reorder columns

head(data_relocate(iris, select = "Species", before = "Sepal.Length"))
head(data_relocate(iris, select = "Species", before = "Sepal.Width"))
head(data_relocate(iris, select = "Sepal.Width", after = "Species"))

# which is same as

head(data_relocate(iris, select = "Sepal.Width", after = -1))

# Reorder multiple columns

head(data_relocate(iris, select = c("Species", "Petal.Length"), after = "Sepal.Width"))

# which is same as

head(data_relocate(iris, select = c("Species", "Petal.Length"), after = 2))

# Reorder columns

head(data_reorder(iris, c("Species", "Sepal.Length")))

# Remove columns

head(data_remove(iris, "Sepal.Length"))
head(data_remove(iris, starts_with("Sepal")))

```
```

data_replicate
Expand (i.e. replicate rows) a data frame

```

\section*{Description}

Expand a data frame by replicating rows based on another variable that contains the counts of replications per row.

\section*{Usage}
```

data_replicate(
data,
expand = NULL,
select = NULL,
exclude = NULL,
remove_na = FALSE,
ignore_case = FALSE,
verbose = TRUE,
regex = FALSE,
)

```

\section*{Arguments}

\section*{data}

A data frame.
expand
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
\begin{tabular}{ll} 
exclude & \begin{tabular}{l} 
See select, however, column names matched by the pattern from exclude will \\
be excluded instead of selected. If NULL (the default), excludes no columns.
\end{tabular} \\
remove_na & \begin{tabular}{l} 
Logical. If TRUE, missing values in the column provided in expand are removed \\
from the data frame. If FALSE and expand contains missing values, the function \\
will throw an error.
\end{tabular} \\
ignore_case & \begin{tabular}{l} 
Logical, if TRUE and when one of the select-helpers or a regular expression is \\
used in select, ignores lower/upper case in the search pattern when matching \\
against variable names.
\end{tabular} \\
verbose & \begin{tabular}{l} 
Toggle warnings.
\end{tabular} \\
regex & \begin{tabular}{l} 
Logical, if TRUE, the search pattern from select will be treated as regular ex- \\
pression. When regex = TRUE, select must be a character string (or a variable \\
containing a character string) and is not allowed to be one of the supported
\end{tabular} \\
select-helpers or a character vector of length > 1. regex = TRUE is compara- \\
ble to using one of the two select-helpers, select = contains("") or select \\
= regex(""), however, since the select-helpers may not work when called from \\
inside other functions (see 'Details'), this argument may be used as workaround.
\end{tabular}

\section*{Value}

A dataframe with each row replicated as many times as defined in expand.

\section*{Examples}
```

data(mtcars)
data_replicate(head(mtcars), "carb")

```
```

data_restoretype Restore the type of columns according to a reference data frame

```

\section*{Description}

Restore the type of columns according to a reference data frame

\section*{Usage \\ data_restoretype(data, reference = NULL, ...)}

\section*{Arguments}
data A data frame for which to restore the column types.
reference A reference data frame from which to find the correct column types. If NULL, each column is converted to numeric if it doesn't generate NAs. For example, \(c(" 1 ", " 2 ")\) can be converted to numeric but not c("Sepal.Length").
... Currently not used.

\section*{Value}

A data frame with columns whose types have been restored based on the reference data frame.

\section*{Examples}
```

data <- data.frame(
Sepal.Length = c("1", "3", "2"),
Species = c("setosa", "versicolor", "setosa"),
New = c("1", "3", "4")
)
fixed <- data_restoretype(data, reference = iris)
summary(fixed)

```
```

data_rotate Rotate a data frame

```

\section*{Description}

This function rotates a data frame, i.e. columns become rows and vice versa. It's the equivalent of using \(t()\) but restores the data.frame class, preserves attributes and prints a warning if the data type is modified (see example).

\section*{Usage}
data_rotate(data, rownames \(=\) NULL, colnames \(=\) FALSE, verbose \(=\) TRUE)
data_transpose(data, rownames \(=\) NULL, colnames \(=\) FALSE, verbose \(=\) TRUE)

\section*{Arguments}
data A data frame.
rownames Character vector (optional). If not NULL, the data frame's rownames will be added as (first) column to the output, with rownames being the name of this column.
colnames Logical or character vector (optional). If TRUE, the values of the first column in \(x\) will be used as column names in the rotated data frame. If a character vector, values from that column are used as column names.
verbose Toggle warnings.

\section*{Value}

A (rotated) data frame.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```

x <- mtcars[1:3, 1:4]
x
data_rotate(x)
data_rotate(x, rownames = "property")

# use values in 1. column as column name

data_rotate(x, colnames = TRUE)
data_rotate(x, rownames = "property", colnames = TRUE)

# use either first column or specific column for column names

x <- data.frame(a = 1:5, b = 11:15, c = 21:25)
data_rotate(x, colnames = TRUE)
data_rotate(x, colnames = "c")

```
data_seek Find variables by their names, variable or value labels

\section*{Description}

This functions seeks variables in a data frame, based on patterns that either match the variable name (column name), variable labels, value labels or factor levels. Matching variable and value labels only works for "labelled" data, i.e. when the variables either have a label attribute or labels attribute.
data_seek() is particular useful for larger data frames with labelled data - finding the correct variable name can be a challenge. This function helps to find the required variables, when only certain patterns of variable names or labels are known.

\section*{Usage}
data_seek(data, pattern, seek = c("names", "labels"), fuzzy = FALSE)

\section*{Arguments}
\[
\begin{array}{ll}
\text { data } & \text { A data frame. } \\
\text { pattern } & \begin{array}{l}
\text { Character string (regular expression) to be matched in data. May also be a char- } \\
\text { acter vector of length > 1. pattern is searched for in column names, variable } \\
\text { label and value labels attributes, or factor levels of variables in data. }
\end{array} \\
\text { seek } & \begin{array}{c}
\text { Character vector, indicating where pattern is sought. Use one or more of the } \\
\text { following options: }
\end{array} \\
\text { - "names": Searches in column names. "column_names" and "columns" are } \\
\text { aliases for "names". } \\
\text { - "labels": Searches in variable labels. Only applies when a label attribute } \\
\text { is set for a variable. } \\
\text { • "values": Searches in value labels or factor levels. Only applies when a } \\
\text { labels attribute is set for a variable, or if a variable is a factor. "levels" } \\
\text { is an alias for "values". } \\
\text { • "all": Searches in all of the above. }
\end{array}
\]

\section*{Value}

A data frame with three columns: the column index, the column name and - if available - the variable label of all matched variables in data.

\section*{Examples}
```


# seek variables with "Length" in variable name or labels

data_seek(iris, "Length")

# seek variables with "dependency" in names or labels

# column "e42dep" has a label-attribute "elder's dependency"

data(efc)
data_seek(efc, "dependency")

# "female" only appears as value label attribute - default search is in

# variable names and labels only, so no match

data_seek(efc, "female")

# when we seek in all sources, we find the variable "e16sex"

data_seek(efc, "female", seek = "all")

# typo, no match

data_seek(iris, "Lenght")

# typo, fuzzy match

data_seek(iris, "Lenght", fuzzy = TRUE)

```
data_select Find or get columns in a data frame based on search patterns

\section*{Description}
extract_column_names() returns column names from a data set that match a certain search pattern, while data_select () returns the found data.

\section*{Usage}
```

    data_select(
    ```
        data,
    select \(=\) NULL,
    exclude \(=\) NULL,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
    )
    get_columns(
        data,
        select \(=\) NULL,
        exclude \(=\) NULL,
        ignore_case = FALSE,
        regex = FALSE,
        verbose \(=\) TRUE,
        )
    extract_column_names(
        data,
        select \(=\) NULL,
        exclude = NULL,
        ignore_case = FALSE,
        regex = FALSE,
        verbose \(=\) TRUE,
        ...
    )
    data_find(
        data,
        select \(=\) NULL,
        exclude = NULL,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE,
```

)
find_columns(
data,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}

\section*{data}
select
exclude

A data frame.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function \((x)\) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a-, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".

See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.
... Arguments passed down to other functions. Mostly not used yet.

\section*{Details}

Specifically for data_select(), select can also be a named character vector. In this case, the names are used to rename the columns in the output data frame. See 'Examples'.
Note that it is possible to either pass an entire select helper or only the pattern inside a select helper as a function argument:
```

foo <- function(data, pattern) {
extract_column_names(data, select = starts_with(pattern))
}
foo(iris, pattern = "Sep")
foo2 <- function(data, pattern) {
extract_column_names(data, select = pattern)
}
foo2(iris, pattern = starts_with("Sep"))

```

This means that it is also possible to use loop values as arguments or patterns:
```

for (i in c("Sepal", "Sp")) {
head(iris) |>
extract_column_names(select = starts_with(i)) |>
print()
}

```

However, this behavior is limited to a "single-level function". It will not work in nested functions, like below:
```

inner <- function(data, arg) {
extract_column_names(data, select = arg)
}
outer <- function(data, arg) {
inner(data, starts_with(arg))
}
outer(iris, "Sep")

```

In this case, it is better to pass the whole select helper as the argument of outer():
```

outer <- function(data, arg) {
inner(data, arg)
}
outer(iris, starts_with("Sep"))

```

\section*{Value}
extract_column_names() returns a character vector with column names that matched the pattern in select and exclude, or NULL if no matching column name was found. data_select () returns a data frame with matching columns.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```


# Find columns names by pattern

extract_column_names(iris, starts_with("Sepal"))
extract_column_names(iris, ends_with("Width"))
extract_column_names(iris, regex("<br>."))
extract_column_names(iris, c("Petal.Width", "Sepal.Length"))

# starts with "Sepal", but not allowed to end with "width"

extract_column_names(iris, starts_with("Sepal"), exclude = contains("Width"))

# find numeric with mean > 3.5

numeric_mean_35 <- function(x) is.numeric(x) \&\& mean(x, na.rm = TRUE) > 3.5
extract_column_names(iris, numeric_mean_35)

# rename returned columns for "data_select()"

head(data_select(mtcars, c(`Miles per Gallon` = "mpg", Cylinders = "cyl")))

```

\section*{Description}

Separates a single variable into multiple new variables.

\section*{Usage}
data_separate( data, select \(=\) NULL, new_columns = NULL, separator = "[^[:alnum:]]+", guess_columns = NULL,
        merge_multiple = FALSE,
        merge_separator = "",
        fill = "right",
        extra = "drop_right",
        convert_na = TRUE,
        exclude = NULL,
        append = FALSE,
        ignore_case = FALSE,
        verbose = TRUE,
        regex = FALSE,
    )

\section*{Arguments}
data
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is. numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
new_columns The names of the new columns, as character vector. If more than one variable was selected (in select), the new names are prefixed with the name of the original column. new_columns can also be a list of (named) character vectors when multiple variables should be separated. See 'Examples'.
separator Separator between columns. Can be a character vector, which is then treated as regular expression, or a numeric vector that indicates at which positions the string values will be split.
guess_columns If new_columns is not given, the required number of new columns is guessed based on the results of value splitting. For example, if a variable is split into three new columns, this will be considered as the required number of new columns, and columns are named "split_1", "split_2" and "split_3". When values from a variable are split into different amount of new columns, the guess_column can be either "mode" (number of new columns is based on the most common number of splits), "min" or "max" to use the minimum resp. maximum number of possible splits as required number of columns.
merge_multiple Logical, if TRUE and more than one variable is selected for separating, new columns can be merged. Value pairs of all split variables are merged.
merge_separator
Separator string when merge_multiple \(=\) TRUE. Defines the string that is used to merge values together.
fill How to deal with values that return fewer new columns after splitting? Can be "left" (fill missing columns from the left with NA), "right" (fill missing columns from the right with NA) or "value_left" or "value_right" to fill missing columns from left or right with the left-most or right-most values.
extra How to deal with values that return too many new columns after splitting? Can be "drop_left" or "drop_right" to drop the left-most or right-most values, or "merge_left" or "merge_right" to merge the left- or right-most value together, and keeping all remaining values as is.
convert_na Logical, if TRUE, character "NA" values are converted into real NA values.
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
\begin{tabular}{ll} 
append & \begin{tabular}{l} 
Logical, if FALSE (default), removes original columns that were separated. If \\
TRUE, all columns are preserved and the new columns are appended to the data \\
frame.
\end{tabular} \\
ignore_case & \begin{tabular}{l} 
Logical, if TRUE and when one of the select-helpers or a regular expression is \\
used in select, ignores lower/upper case in the search pattern when matching \\
against variable names.
\end{tabular} \\
verbose & \begin{tabular}{l} 
Toggle warnings.
\end{tabular} \\
regex & \begin{tabular}{l} 
Logical, if TRUE, the search pattern from select will be treated as regular ex- \\
pression. When regex = TRUE, select must be a character string (or a variable \\
containing a character string) and is not allowed to be one of the supported \\
select-helpers or a character vector of length > 1. regex = TRUE is compara- \\
ble to using one of the two select-helpers, select = contains("") or select \\
= regex(""), however, since the select-helpers may not work when called from \\
inside other functions (see 'Details'), this argument may be used as workaround.
\end{tabular} \\
\(\ldots\) & Currently not used.
\end{tabular}

\section*{Value}

A data frame with the newly created variable(s), or - when append = TRUE - data including new variables.

\section*{See Also}
```

data_unite()

```

\section*{Examples}
```


# simple case

d <- data.frame(
x = c("1.a.6", "2.b.7", "3.c.8"),
stringsAsFactors = FALSE
)
d
data_separate(d, new_columns = c("a", "b", "c"))

# guess number of columns

d <- data.frame(
x = c("1.a.6", NA, "2.b.6.7", "3.c", "x.y.z"),
stringsAsFactors = FALSE
)
d
data_separate(d, guess_columns = "mode")
data_separate(d, guess_columns = "max")

# drop left-most column

data_separate(d, guess_columns = "mode", extra = "drop_left")

# merge right-most column

data_separate(d, guess_columns = "mode", extra = "merge_right")

```
```


# fill columns with fewer values with left-most values

data_separate(d, guess_columns = "mode", fill = "value_left")

# fill and merge

data_separate(
d,
guess_columns = "mode",
fill = "value_left",
extra = "merge_right"
)

# multiple columns to split

d <- data.frame(
x = c("1.a.6", "2.b.7", "3.c.8"),
y = c("x.y.z", "10.11.12", "m.n.o"),
stringsAsFactors = FALSE
)
d

# split two columns, default column names

data_separate(d, guess_columns = "mode")

# split into new named columns, repeating column names

data_separate(d, new_columns = c("a", "b", "c"))

# split selected variable new columns

data_separate(d, select = "y", new_columns = c("a", "b", "c"))

# merge multiple split columns

data_separate(
d,
new_columns = c("a", "b", "c"),
merge_multiple = TRUE
)

# merge multiple split columns

data_separate(
d,
new_columns = c("a", "b", "c"),
merge_multiple = TRUE,
merge_separator = "-"
)

# separate multiple columns, give proper column names

d_sep <- data.frame(
x = c("1.a.6", "2.b.7.d", "3.c.8", "5.j"),
y = c("m.n.99.22", "77.f.g.34", "44.9", NA),
stringsAsFactors = FALSE
)
data_separate(
d_sep,
select = c("x", "y"),

```
```

    new_columns = list(
        x = c("A", "B", "C"), # separate "x" into three columns
        y = c("EE", "FF", "GG", "HH") # separate "y" into four columns
    ),
    verbose = FALSE
    )

```
data_summary Summarize data

\section*{Description}

This function can be used to compute summary statistics for a data frame or a matrix.

\section*{Usage}
```

data_summary (x, ...)
\#\# S3 method for class 'data.frame'
data_summary ( $x, \ldots$, by $=$ NULL, remove_na $=$ FALSE)

```

\section*{Arguments}
x
... One or more named expressions that define the new variable name and the function to compute the summary statistic. Example: mean_sepal_width = mean(Sepal.Width). The expression can also be provided as a character string, e.g. "mean_sepal_width = mean(Sepal.Width)". The summary function n() can be used to count the number of observations.
by Optional character string, indicating the name of a variable in x . If supplied, the data will be split by this variable and summary statistics will be computed for each group.
remove_na Logical. If TRUE, missing values are omitted from the grouping variable. If FALSE (default), missing values are included as a level in the grouping variable.

\section*{Value}

A data frame with the requested summary statistics.

\section*{Examples}
```

data(iris)
data_summary(iris, MW = mean(Sepal.Width), SD = sd(Sepal.Width))
data_summary(
iris,
MW = mean(Sepal.Width),
SD = sd(Sepal.Width),
by = "Species"

```
)
```

    # same as
    d <- data_group(iris, "Species")
    data_summary(d, MW = mean(Sepal.Width), SD = sd(Sepal.Width))
    # multiple groups
    data(mtcars)
    data_summary(mtcars, MW = mean(mpg), SD = sd(mpg), by = c("am", "gear"))
    # expressions can also be supplied as character strings
    data_summary(mtcars, "MW = mean(mpg)", "SD = sd(mpg)", by = c("am", "gear"))
    # count observations within groups
    data_summary(mtcars, observations = n(), by = c("am", "gear"))
    # first and last observations of "mpg" within groups
    data_summary(
        mtcars,
    first = mpg[1],
    last = mpg[length(mpg)],
    by = c("am", "gear")
    )

```
data_tabulate

Create frequency and crosstables of variables

\section*{Description}

This function creates frequency or crosstables of variables, including the number of levels/values as well as the distribution of raw, valid and cumulative percentages. For crosstables, row, column and cell percentages can be calculated.
```

Usage
data_tabulate(x, ...)
\#\# Default S3 method:
data_tabulate(
x,
by = NULL,
drop_levels = FALSE,
weights = NULL,
remove_na = FALSE,
proportions = NULL,
name = NULL,
verbose = TRUE,
)

```
```


## S3 method for class 'data.frame'

data_tabulate(
x,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
by = NULL,
drop_levels = FALSE,
weights = NULL,
remove_na = FALSE,
proportions = NULL,
collapse = FALSE,
verbose = TRUE,
)

## S3 method for class 'datawizard_tables'

as.data.frame(
x,
row.names = NULL,
optional = FALSE,
...,
stringsAsFactors = FALSE,
add_total = FALSE
)

```

\section*{Arguments}

\section*{X}
... not used.
by Optional vector or factor. If supplied, a crosstable is created. If \(x\) is a data frame, by can also be a character string indicating the name of a variable in \(x\).
drop_levels Logical, if FALSE, factor levels that do not occur in the data are included in the table (with frequency of zero), else unused factor levels are dropped from the frequency table.
weights Optional numeric vector of weights. Must be of the same length as \(x\). If weights is supplied, weighted frequencies are calculated.
remove_na Logical, if FALSE, missing values are included in the frequency or crosstable, else missing values are omitted.
proportions Optional character string, indicating the type of percentages to be calculated. Only applies to crosstables, i.e. when by is not NULL. Can be "row" (row percentages), "column" (column percentages) or "full" (to calculate relative frequencies for the full table).
name Optional character string, which includes the name that is used for printing.

```

optional logical. If TRUE, setting row names and converting column names (to syntac-
tic names: see make.names) is optional. Note that all of R's base package
as.data.frame() methods use optional only for column names treatment, ba-
sically with the meaning of data.frame(*, check.names = !optional). See
also the make.names argument of the matrix method.
stringsAsFactors
logical: should the character vector be converted to a factor?
add_total For crosstables (i.e. when by is not NULL), a row and column with the total N val-
ues are added to the data frame. add_total has no effect in as.data.frame()
for simple frequency tables.

```

\section*{Details}

There is an as.data.frame() method, to return the frequency tables as a data frame. The structure of the returned object is a nested data frame, where the first column contains name of the variable for which frequencies were calculated, and the second column is a list column that contains the frequency tables as data frame. See 'Examples'.

\section*{Value}

A data frame, or a list of data frames, with one frequency table as data frame per variable.

\section*{Crosstables}

If by is supplied, a crosstable is created. The crosstable includes <NA> (missing) values by default. The first column indicates values of \(x\), the first row indicates values of by (including missing values). The last row and column contain the total frequencies for each row and column, respectively. Setting remove_na = FALSE will omit missing values from the crosstable. Setting proportions to "row" or "column" will add row or column percentages. Setting proportions to "full" will add relative frequencies for the full table.

\section*{Note}

There are print_html() and print_md() methods available for printing frequency or crosstables in HTML and markdown format, e.g. print_html(data_tabulate(x)).

\section*{Examples}
```


# frequency tables -------

# ------------------------

data(efc)

# vector/factor

data_tabulate(efc\$c172code)

# drop missing values

data_tabulate(efc\$c172code, remove_na = TRUE)

# data frame

data_tabulate(efc, c("e42dep", "c172code"))

```
```


# grouped data frame

suppressPackageStartupMessages(library(poorman, quietly = TRUE))
efc %>%
group_by(c172code) %>%
data_tabulate("e16sex")

# collapse tables

efc %>%
group_by(c172code) %>%
data_tabulate("e16sex", collapse = TRUE)

# for larger N's (> 100000), a big mark is automatically added

set.seed(123)
x <- sample(1:3, 1e6, TRUE)
data_tabulate(x, name = "Large Number")

# to remove the big mark, use "print(..., big_mark = "")"

print(data_tabulate(x), big_mark = "")

# weighted frequencies

set.seed(123)
efc$weights <- abs(rnorm(n = nrow(efc), mean = 1, sd = 0.5))
data_tabulate(efc$e42dep, weights = efc\$weights)

# crosstables ------

# -----------------

# add some missing values

set.seed(123)
efc\$e16sex[sample.int(nrow(efc), 5)] <- NA
data_tabulate(efc, "c172code", by = "e16sex")

# add row and column percentages

data_tabulate(efc, "c172code", by = "e16sex", proportions = "row")
data_tabulate(efc, "c172code", by = "e16sex", proportions = "column")

# omit missing values

data_tabulate(
efc$c172code,
    by = efc$e16sex,
proportions = "column",
remove_na = TRUE
)

# round percentages

out <- data_tabulate(efc, "c172code", by = "e16sex", proportions = "column")
print(out, digits = 0)

# coerce to data frames

result <- data_tabulate(efc, "c172code", by = "e16sex")
as.data.frame(result)

```
```

as.data.frame(result)$table
as.data.frame(result, add_total = TRUE)$table

```
data_to_long Reshape (pivot) data from wide to long

\section*{Description}

This function "lengthens" data, increasing the number of rows and decreasing the number of columns.
This is a dependency-free base-R equivalent of tidyr: :pivot_longer().

\section*{Usage}
```

data_to_long(
data,
select = "all",
names_to = "name",
names_prefix = NULL,
names_sep = NULL,
names_pattern = NULL,
values_to = "value",
values_drop_na = FALSE,
rows_to = NULL,
ignore_case = FALSE,
regex = FALSE,
...,
cols
)
reshape_longer(
data,
select = "all",
names_to = "name",
names_prefix = NULL,
names_sep = NULL,
names_pattern = NULL,
values_to = "value",
values_drop_na = FALSE,
rows_to = NULL,
ignore_case = FALSE,
regex = FALSE,
...,
cols
)

```

\section*{Arguments}
data A data frame to convert to long format, so that it has more rows and fewer columns after the operation.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function (x) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
names_to The name of the new column (variable) that will contain the names from columns in select as values, to identify the source of the values. names_to can be a character vector with more than one column name, in which case names_sep or names_pattern must be provided in order to identify which parts of the column names go into newly created columns. See also 'Examples'.
names_prefix A regular expression used to remove matching text from the start of each variable name.
names_sep, names_pattern
If names_to contains multiple values, this argument controls how the column name is broken up. names_pattern takes a regular expression containing matching groups, i.e. "()".
values_to The name of the new column that will contain the values of the columns in select.
values_drop_na If TRUE, will drop rows that contain only NA in the values_to column. This effectively converts explicit missing values to implicit missing values, and should generally be used only when missing values in data were created by its structure.
\[
\begin{array}{ll}
\text { rows_to } & \begin{array}{l}
\text { The name of the column that will contain the row names or row numbers from } \\
\text { the original data. If NULL, will be removed. } \\
\text { ignore_case } \\
\text { Logical, if TRUE and when one of the select-helpers or a regular expression is } \\
\text { used in select, ignores lower/upper case in the search pattern when matching } \\
\text { against variable names. }
\end{array} \\
\text { regex } & \begin{array}{l}
\text { Logical, if TRUE, the search pattern from select will be treated as regular ex- } \\
\text { pression. When regex = TRUE, select must be a character string (or a variable } \\
\text { containing a character string) and is not allowed to be one of the supported } \\
\text { select-helpers or a character vector of length > 1. regex = TRUE is compara- } \\
\text { ble to using one of the two select-helpers, select = contains("") or select } \\
\text { = regex(""), however, since the select-helpers may not work when called from } \\
\text { inside other functions (see 'Details'), this argument may be used as workaround. }
\end{array} \\
\ldots & \begin{array}{l}
\text { Currently not used. }
\end{array} \\
\text { cols } & \begin{array}{l}
\text { Identical to select. This argument is here to ensure compatibility with tidyr: :pivot_longer(). } \\
\text { If both select and cols are provided, cols is used. }
\end{array}
\end{array}
\]

\section*{Details}

Reshaping data into long format usually means that the input data frame is in wide format, where multiple measurements taken on the same subject are stored in multiple columns (variables). The long format stores the same information in a single column, with each measurement per subject stored in a separate row. The values of all variables that are not in select will be repeated.

The necessary information for data_to_long() is:
- The columns that contain the repeated measurements (select).
- The name of the newly created column that will contain the names of the columns in select (names_to), to identify the source of the values. names_to can also be a character vector with more than one column name, in which case names_sep or names_pattern must be provided to specify which parts of the column names go into the newly created columns.
- The name of the newly created column that contains the values of the columns in select (values_to).

In other words: repeated measurements that are spread across several columns will be gathered into a single column (values_to), with the original column names, that identify the source of the gathered values, stored in one or more new columns (names_to).

\section*{Value}

If a tibble was provided as input, reshape_longer() also returns a tibble. Otherwise, it returns a data frame.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```

wide_data <- setNames(
data.frame(replicate(2, rnorm(8))),
c("Time1", "Time2")
)
wide_data\$ID <- 1:8
wide_data

# Default behaviour (equivalent to tidyr::pivot_longer(wide_data, cols = 1:3))

# probably doesn't make much sense to mix "time" and "id"

data_to_long(wide_data)

# Customizing the names

data_to_long(
wide_data,
select = c("Time1", "Time2"),
names_to = "Timepoint",
values_to = "Score"
)

# Reshape multiple columns into long format.

mydat <- data.frame(
age = c(20, 30, 40),
sex = c("Female", "Male", "Male"),
score_t1 = c(30, 35, 32),
score_t2 = c (33, 34, 37),
score_t3 = c(36, 35, 38),
speed_t1 = c(2, 3, 1),
speed_t2 = c(3, 4, 5),
speed_t3 = c(1, 8, 6)
)

# The column names are split into two columns: "type" and "time". The

# pattern for splitting column names is provided in `names_pattern`. Values

# of all "score_*" and "speed_*" columns are gathered into a single column

# named "count".

data_to_long(
mydat,
select = 3:8,
names_to = c("type", "time"),
names_pattern = "(score|speed)_t(<br>d+)",
values_to = "count"
)

# Full example

# ------------------

```
```

data_to_wide

```
```

data <- psych::bfi \# Wide format with one row per participant's personality test

```
data <- psych::bfi # Wide format with one row per participant's personality test
# Pivot long format
# Pivot long format
very_long_data <- data_to_long(data,
very_long_data <- data_to_long(data,
        select = regex("\\d"), # Select all columns that contain a digit
        select = regex("\\d"), # Select all columns that contain a digit
        names_to = "Item",
        names_to = "Item",
        values_to = "Score",
        values_to = "Score",
        rows_to = "Participant"
        rows_to = "Participant"
)
)
head(very_long_data)
head(very_long_data)
even_longer_data <- data_to_long(
even_longer_data <- data_to_long(
        tidyr::who,
        tidyr::who,
        select = new_sp_m014:newrel_f65,
        select = new_sp_m014:newrel_f65,
        names_to = c("diagnosis", "gender", "age"),
        names_to = c("diagnosis", "gender", "age"),
        names_pattern = "new_?(.*)_(.)(.*)",
        names_pattern = "new_?(.*)_(.)(.*)",
        values_to = "count"
        values_to = "count"
)
)
head(even_longer_data)
```

head(even_longer_data)

```
data_to_wide Reshape (pivot) data from long to wide

\section*{Description}

This function "widens" data, increasing the number of columns and decreasing the number of rows. This is a dependency-free base-R equivalent of tidyr: : pivot_wider().

\section*{Usage}
```

data_to_wide(
data,
id_cols = NULL,
values_from = "Value",
names_from = "Name",
names_sep = "_",
names_prefix = "",
names_glue = NULL,
values_fill = NULL,
verbose = TRUE,
)
reshape_wider(
data,
id_cols = NULL,
values_from = "Value",
names_from = "Name",

```
```

    names_sep = "_",
    names_prefix = "",
    names_glue = NULL,
    values_fill = NULL,
    verbose = TRUE,
    )

```

\section*{Arguments}
\begin{tabular}{|c|c|}
\hline data & A data frame to convert to wide format, so that it has more columns and fewer rows post-widening than pre-widening. \\
\hline id_cols & The name of the column that identifies the rows in the data by which observations are grouped and the gathered data is spread into new columns. Usually, this is a variable containing an ID for observations that have been repeatedly measured. If NULL, it will use all remaining columns that are not in names_from or values_from as ID columns. id_cols can also be a character vector with more than one name of identifier columns. See also 'Details' and 'Examples'. \\
\hline values_from & The name of the columns in the original data that contains the values used to fill the new columns created in the widened data. \\
\hline names_from & The name of the column in the original data whose values will be used for naming the new columns created in the widened data. Each unique value in this column will become the name of one of these new columns. In case names_prefix is provided, column names will be concatenated with the string given in names_prefix. \\
\hline names_sep & If names_from or values_from contains multiple variables, this will be used to join their values together into a single string to use as a column name. \\
\hline names_prefix & String added to the start of every variable name. This is particularly useful if names_from is a numeric vector and you want to create syntactic variable names. \\
\hline names_glue & Instead of names_sep and names_prefix, you can supply a glue specification that uses the names_from columns to create custom column names. Note that the only delimiters supported by names_glue are curly brackets, \{ and \}. \\
\hline values_fill & Optionally, a (scalar) value that will be used to replace missing values in the new columns created. \\
\hline verbose & Toggle warnings. \\
\hline & Not used for now. \\
\hline
\end{tabular}

\section*{Details}

Reshaping data into wide format usually means that the input data frame is in long format, where multiple measurements taken on the same subject are stored in multiple rows. The wide format stores the same information in a single row, with each measurement stored in a separate column. Thus, the necessary information for data_to_wide() is:
- The name of the column(s) that identify the groups or repeated measurements (id_cols).
- The name of the column whose values will become the new column names (names_from). Since these values may not necessarily reflect appropriate column names, you can use names_prefix to add a prefix to each newly created column name.
- The name of the column that contains the values (values_from) for the new columns that are created by names_from.

In other words: repeated measurements, as indicated by id_cols, that are saved into the column values_from will be spread into new columns, which will be named after the values in names_from. See also 'Examples'.

\section*{Value}

If a tibble was provided as input, data_to_wide() also returns a tibble. Otherwise, it returns a data frame.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder (), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```

data_long <- read.table(header = TRUE, text = "
subject sex condition measurement

| 1 | M | control | 7.9 |
| :--- | :--- | ---: | ---: |
| 1 | M | cond1 | 12.3 |
| 1 | M | cond2 | 10.7 |
| 2 | F | control | 6.3 |
| 2 | F | cond1 | 10.6 |
| 2 | F | cond2 | 11.1 |
| 3 | F | control | 9.5 |
| 3 | F | cond1 | 13.1 |
| 3 | F | cond2 | 13.8 |
| 4 | M | control | 11.5 |
| 4 | M | cond1 | 13.4 |
| 4 | M | cond2 | $12.9^{\prime \prime}$ ) |

# converting long data into wide format

data_to_wide(
data_long,
id_cols = "subject",
names_from = "condition",

```
```

    values_from = "measurement"
    )

# converting long data into wide format with custom column names

data_to_wide(
data_long,
id_cols = "subject",
names_from = "condition",
values_from = "measurement",
names_prefix = "Var.",
names_sep = "."
)

# converting long data into wide format, combining multiple columns

production <- expand.grid(
product = c("A", "B"),
country = c("AI", "EI"),
year = 2000:2014
)
production <- data_filter(production, (product == "A" \& country == "AI") | product == "B")
production\$production <- rnorm(nrow(production))
data_to_wide(
production,
names_from = c("product", "country"),
values_from = "production",
names_glue = "prod_{product}_{country}"
)

# using the "sleepstudy" dataset

data(sleepstudy, package = "lme4")

# the sleepstudy data contains repeated measurements of average reaction

# times for each subjects over multiple days, in a sleep deprivation study.

# It is in long-format, i.e. each row corresponds to a single measurement.

# The variable "Days" contains the timepoint of the measurement, and

# "Reaction" contains the measurement itself. Converting this data to wide

# format will create a new column for each day, with the reaction time as the

# value.

head(sleepstudy)
data_to_wide(
sleepstudy,
id_cols = "Subject",
names_from = "Days",
values_from = "Reaction"
)

# clearer column names

data_to_wide(
sleepstudy,
id_cols = "Subject",
names_from = "Days",

```
```

    values_from = "Reaction",
    names_prefix = "Reaction_Day_"
    )

# For unequal group sizes, missing information is filled with NA

d <- subset(sleepstudy, Days %in% c(0, 1, 2, 3, 4))[c(1:9, 11:13, 16:17, 21), ]

# long format, different number of "Subjects"

d
data_to_wide(
d,
id_cols = "Subject",
names_from = "Days",
values_from = "Reaction",
names_prefix = "Reaction_Day_"
)

# filling missing values with 0

data_to_wide(
d,
id_cols = "Subject",
names_from = "Days",
values_from = "Reaction",
names_prefix = "Reaction_Day_",
values_fill = 0
)

```
data_unique Keep only one row from all with duplicated IDs

\section*{Description}

From all rows with at least one duplicated ID, keep only one. Methods for selecting the duplicated row are either the first duplicate, the last duplicate, or the "best" duplicate (default), based on the duplicate with the smallest number of NA. In case of ties, it picks the first duplicate, as it is the one most likely to be valid and authentic, given practice effects.

Contrarily to dplyr:: distinct(), data_unique() keeps all columns.

\section*{Usage}
```

    data_unique(
        data,
    select = NULL,
    keep = "best",
    exclude = NULL,
    ignore_case = FALSE,
    regex = FALSE,
    ```
```

    verbose = TRUE
    )

```

\section*{Arguments}

verbose Toggle warnings.

\section*{Value}

A data frame, containing only the chosen duplicates.

\section*{See Also}
```

data_duplicated()

```

\section*{Examples}
```

df1 <- data.frame(
id = c(1, 2, 3, 1, 3),
item1 = c(NA, 1, 1, 2, 3),
item2 = c(NA, 1, 1, 2, 3),
item3 = c(NA, 1, 1, 2, 3)
)
data_unique(df1, select = "id")

```
data_unite Unite ("merge") multiple variables

\section*{Description}

Merge values of multiple variables per observation into one new variable.

\section*{Usage}
```

    data_unite(
        data,
        new_column = NULL,
        select = NULL,
        exclude = NULL,
        separator = "_",
        append = FALSE,
        remove_na = FALSE,
        ignore_case = FALSE,
        verbose = TRUE,
        regex = FALSE,
    )
    ```

\section*{Arguments}

select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select \(=\operatorname{regex}("\) "), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
... Currently not used.

\section*{Value}
data, with a newly created variable.

\section*{See Also}
```

data_separate()

```

\section*{Examples}
```

d <- data.frame(
x = 1:3,
y = letters[1:3],
z = 6:8
)
d
data_unite(d, new_column = "xyz")
data_unite(d, new_column = "xyz", remove = FALSE)
data_unite(d, new_column = "xyz", select = c("x", "z"))
data_unite(d, new_column = "xyz", select = c("x", "z"), append = TRUE)

```
demean Compute group-meaned and de-meaned variables

\section*{Description}
demean() computes group- and de-meaned versions of a variable that can be used in regression analysis to model the between- and within-subject effect. degroup() is more generic in terms of the centering-operation. While demean() always uses mean-centering, degroup() can also use the mode or median for centering.

\section*{Usage}
demean(
\(x\),
select,
by,
suffix_demean = "_within",
suffix_groupmean = "_between",
add_attributes = TRUE,
verbose = TRUE,
group \(=\) NULL
```

)
degroup(
x,
select,
by,
center = "mean",
suffix_demean = "_within",
suffix_groupmean = "_between",
add_attributes = TRUE,
verbose = TRUE,
group = NULL
)
detrend(
x,
select,
by,
center = "mean",
suffix_demean = "_within",
suffix_groupmean = "_between",
add_attributes = TRUE,
verbose = TRUE,
group = NULL
)

```

\section*{Arguments}
x
select Character vector (or formula) with names of variables to select that should be group- and de-meaned.
by Character vector (or formula) with the name of the variable(s) that indicates the group- or cluster-ID. For cross-classified designs, by can also identify two or more variables as group- or cluster-IDs. See also section De-meaning for crossclassified designs below.
suffix_demean, suffix_groupmean
String value, will be appended to the names of the group-meaned and de-meaned variables of \(x\). By default, de-meaned variables will be suffixed with "_within" and grouped-meaned variables with "_between".
add_attributes Logical, if TRUE, the returned variables gain attributes to indicate the within- and between-effects. This is only relevant when printing model_parameters() - in such cases, the within- and between-effects are printed in separated blocks.
verbose Toggle warnings and messages.
group Deprecated. Use by instead.
center Method for centering. demean() always performs mean-centering, while degroup() can use center = "median" or center = "mode" for median- or mode-centering, and also "min" or "max".

\section*{Value}

A data frame with the group-/de-meaned variables, which get the suffix "_between" (for the groupmeaned variable) and "_wi thin" (for the de-meaned variable) by default.

\section*{Heterogeneity Bias}

Mixed models include different levels of sources of variability, i.e. error terms at each level. When macro-indicators (or level-2 predictors, or higher-level units, or more general: group-level predictors that vary within and across groups) are included as fixed effects (i.e. treated as covariate at level-1), the variance that is left unaccounted for this covariate will be absorbed into the error terms of level-1 and level-2 (Bafumi and Gelman 2006; Gelman and Hill 2007, Chapter 12.6.): "Such covariates contain two parts: one that is specific to the higher-level entity that does not vary between occasions, and one that represents the difference between occasions, within higher-level entities" (Bell et al. 2015). Hence, the error terms will be correlated with the covariate, which violates one of the assumptions of mixed models (iid, independent and identically distributed error terms). This bias is also called the heterogeneity bias (Bell et al. 2015). To resolve this problem, level-2 predictors used as (level-1) covariates should be separated into their "within" and "between" effects by "de-meaning" and "group-meaning": After demeaning time-varying predictors, "at the higher level, the mean term is no longer constrained by Level 1 effects, so it is free to account for all the higher-level variance associated with that variable" (Bell et al. 2015).

\section*{Panel data and correlating fixed and group effects}
demean() is intended to create group- and de-meaned variables for panel regression models (fixed effects models), or for complex random-effect-within-between models (see Bell et al. 2015, 2018), where group-effects (random effects) and fixed effects correlate (see Bafumi and Gelman 2006). This can happen, for instance, when analyzing panel data, which can lead to Heterogeneity Bias. To control for correlating predictors and group effects, it is recommended to include the groupmeaned and de-meaned version of time-varying covariates (and group-meaned version of timeinvariant covariates that are on a higher level, e.g. level-2 predictors) in the model. By this, one can fit complex multilevel models for panel data, including time-varying predictors, time-invariant predictors and random effects.

\section*{Why mixed models are preferred over fixed effects models}

A mixed models approach can model the causes of endogeneity explicitly by including the (separated) within- and between-effects of time-varying fixed effects and including time-constant fixed effects. Furthermore, mixed models also include random effects, thus a mixed models approach is superior to classic fixed-effects models, which lack information of variation in the group-effects or between-subject effects. Furthermore, fixed effects regression cannot include random slopes, which means that fixed effects regressions are neglecting "cross-cluster differences in the effects of lowerlevel controls (which) reduces the precision of estimated context effects, resulting in unnecessarily wide confidence intervals and low statistical power" (Heisig et al. 2017).

\section*{Terminology}

The group-meaned variable is simply the mean of an independent variable within each group (or idlevel or cluster) represented by by. It represents the cluster-mean of an independent variable. The regression coefficient of a group-meaned variable is the between-subject-effect. The de-meaned
variable is then the centered version of the group-meaned variable. De-meaning is sometimes also called person-mean centering or centering within clusters. The regression coefficient of a de-meaned variable represents the within-subject-effect.

\section*{De-meaning with continuous predictors}

For continuous time-varying predictors, the recommendation is to include both their de-meaned and group-meaned versions as fixed effects, but not the raw (untransformed) time-varying predictors themselves. The de-meaned predictor should also be included as random effect (random slope). In regression models, the coefficient of the de-meaned predictors indicates the within-subject effect, while the coefficient of the group-meaned predictor indicates the between-subject effect.

\section*{De-meaning with binary predictors}

For binary time-varying predictors, there are two recommendations. First is to include the raw (untransformed) binary predictor as fixed effect only and the de-meaned variable as random effect (random slope). The alternative would be to add the de-meaned version(s) of binary time-varying covariates as additional fixed effect as well (instead of adding it as random slope). Centering timevarying binary variables to obtain within-effects (level 1) isn't necessary. They have a sensible interpretation when left in the typical \(0 / 1\) format (Hoffmann 2015, chapter 8-2.I). demean() will thus coerce categorical time-varying predictors to numeric to compute the de- and group-meaned versions for these variables, where the raw (untransformed) binary predictor and the de-meaned version should be added to the model.

\section*{De-meaning of factors with more than \(\mathbf{2}\) levels}

Factors with more than two levels are demeaned in two ways: first, these are also converted to numeric and de-meaned; second, dummy variables are created (binary, with \(0 / 1\) coding for each level) and these binary dummy-variables are de-meaned in the same way (as described above). Packages like panelr internally convert factors to dummies before demeaning, so this behaviour can be mimicked here.

\section*{De-meaning interaction terms}

There are multiple ways to deal with interaction terms of within- and between-effects.
- A classical approach is to simply use the product term of the de-meaned variables (i.e. introducing the de-meaned variables as interaction term in the model formula, e.g. \(\mathrm{y} \sim \mathrm{x}\) _within * time_within). This approach, however, might be subject to bias (see Giesselmann \& Schmidt-Catran 2020).
- Another option is to first calculate the product term and then apply the de-meaning to it. This approach produces an estimator "that reflects unit-level differences of interacted variables whose moderators vary within units", which is desirable if no within interaction of two timedependent variables is required. This is what demean() does internally when select contains interaction terms.
- A third option, when the interaction should result in a genuine within estimator, is to "double de-mean" the interaction terms (Giesselmann \& Schmidt-Catran 2018), however, this is currently not supported by demean(). If this is required, the wmb() function from the panelr package should be used.

To de-mean interaction terms for within-between models, simply specify the term as interaction for the select-argument, e.g. select = "a*b" (see 'Examples').

\section*{De-meaning for cross-classified designs}
demean () can also handle cross-classified designs, where the data has two or more groups at the higher (i.e. second) level. In such cases, the by-argument can identify two or more variables that represent the cross-classified group- or cluster-IDs. The de-meaned variables for cross-classified designs are simply subtracting all group means from each individual value, i.e. fully cluster-meancentering (see Guo et al. 2024 for details). Note that de-meaning for cross-classified designs is not equivalent to de-meaning of nested data structures from models with three or more levels, i.e. de-meaning is supposed to work for models like \(\mathrm{y} \sim \mathrm{x}+(1 \mid\) group1 \()+(1 \mid\) group2 \()\), but not for models like \(y \sim x+(1 \mid\) group1/group2).

\section*{Analysing panel data with mixed models using lme4}

A description of how to translate the formulas described in Bell et al. 2018 into R using lmer () from lme 4 can be found in this vignette.

\section*{References}
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- Hoffman L. 2015. Longitudinal analysis: modeling within-person fluctuation and change. New York: Routledge

\section*{See Also}

If grand-mean centering (instead of centering within-clusters) is required, see center(). See performance: :check_heterogeneity_bias() to check for heterogeneity bias.

\section*{Examples}
```

data(iris)
iris$ID <- sample(1:4, nrow(iris), replace = TRUE) # fake-ID
iris$binary <- as.factor(rbinom(150, 1, .35)) \# binary variable
x <- demean(iris, select = c("Sepal.Length", "Petal.Length"), by = "ID")
head(x)
x <- demean(iris, select = c("Sepal.Length", "binary", "Species"), by = "ID")
head(x)

# demean interaction term x*y

dat <- data.frame(
a = c(1, 2, 3, 4, 1, 2, 3, 4),
x = c(4, 3, 3, 4, 1, 2, 1, 2),
y = c(1, 2, 1, 2, 4, 3, 2, 1),
ID = c(1, 2, 3, 1, 2, 3, 1, 2)
)
demean(dat, select = c("a", "x*y"), by = "ID")

# or in formula-notation

demean(dat, select = ~ a + x * y, by = ~ID)

```
    describe_distribution Describe a distribution

\section*{Description}

This function describes a distribution by a set of indices (e.g., measures of centrality, dispersion, range, skewness, kurtosis).

\section*{Usage}
```

describe_distribution(x, ...)

## S3 method for class 'numeric'

describe_distribution(
x,
centrality = "mean",
dispersion = TRUE,
iqr = TRUE,
range = TRUE,
quartiles = FALSE,
ci = NULL,
iterations = 100,
threshold = 0.1,

```
```

    verbose = TRUE,
    )

## S3 method for class 'factor'

describe_distribution(x, dispersion = TRUE, range = TRUE, verbose = TRUE, ...)

## S3 method for class 'data.frame'

describe_distribution(
x,
select = NULL,
exclude = NULL,
centrality = "mean",
dispersion = TRUE,
iqr = TRUE,
range = TRUE,
quartiles = FALSE,
include_factors = FALSE,
ci = NULL,
iterations = 100,
threshold = 0.1,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}
x
...
centrality
dispersion Logical, if TRUE, computes indices of dispersion related to the estimate(s) (SD
iqr Logical, if TRUE, the interquartile range is calculated (based on stats: : \(\operatorname{IQR}()\), using type \(=6\) ).
range \(\quad\) Return the range (min and max).
quartiles
ci
iterations
A numeric vector, a character vector, a data frame, or a list. See Details.
Additional arguments to be passed to or from methods.
The point-estimates (centrality indices) to compute. Character (vector) or list with one or more of these options: "median", "mean", "MAP" (see map_estimate()), "trimmed" (which is just mean (x, trim = threshold)), "mode" or "all". and MAD for mean and median, respectively). Dispersion is not available for "MAP" or "mode" centrality indices.

Return the first and third quartiles (25th and 75pth percentiles).
Confidence Interval (CI) level. Default is NULL, i.e. no confidence intervals are computed. If not NULL, confidence intervals are based on bootstrap replicates (see iterations). If centrality = "all", the bootstrapped confidence interval refers to the first centrality index (which is typically the median).

The number of bootstrap replicates for computing confidence intervals. Only applies when ci is not NULL.


\section*{Details}

If \(x\) is a data frame, only numeric variables are kept and will be displayed in the summary.
If \(x\) is a list, the behavior is different whether \(x\) is a stored list. If \(x\) is stored (for example, describe_distribution(mylist) where mylist was created before), artificial variable names
are used in the summary (Var_1, Var_2, etc.). If \(x\) is an unstored list (for example, describe_distribution(list (mtcars \(\$ r\) then "mtcars\$mpg" is used as variable name.

\section*{Value}

A data frame with columns that describe the properties of the variables.

\section*{Note}

There is also a plot()-method implemented in the see-package.

\section*{Examples}
```

describe_distribution(rnorm(100))
data(iris)
describe_distribution(iris)
describe_distribution(iris, include_factors = TRUE, quartiles = TRUE)
describe_distribution(list(mtcars$mpg, mtcars$cyl))

```
```

distribution_mode Compute mode for a statistical distribution

```

\section*{Description}

Compute mode for a statistical distribution

\section*{Usage}
distribution_mode(x)

\section*{Arguments}
x An atomic vector, a list, or a data frame.

\section*{Value}

The value that appears most frequently in the provided data. The returned data structure will be the same as the entered one.

\section*{See Also}

For continuous variables, the Highest Maximum a Posteriori probability estimate (MAP) may be a more useful way to estimate the most commonly-observed value than the mode. See bayestestR: :map_estimate().

\section*{Examples}
```

    distribution_mode(c(1, 2, 3, 3, 4, 5))
    distribution_mode(c(1.5, 2.3, 3.7, 3.7, 4.0, 5))
    ```
```

efc Sample dataset from the EFC Survey

```

\section*{Description}

Selected variables from the EUROFAMCARE survey. Useful when testing on "real-life" data sets, including random missing values. This data set also has value and variable label attributes.
```

    labels_to_levels Convert value labels into factor levels
    ```

\section*{Description}

\section*{Convert value labels into factor levels}

\section*{Usage}
```

labels_to_levels(x, ...)

## S3 method for class 'factor'

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

## S3 method for class 'data.frame'

labels_to_levels(
x,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
append = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}
x
...
verbose
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric () (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is. numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to
exclude=ends_with("Length") (no -). In case negation should not work For instance, select=-ends_with("Length") (with -) is equivalent to
exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
append
A data frame or factor. Other variable types (e.g. numerics) are not allowed.
Currently not used.
Toggle warnings.

Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in \(x\) will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Details}
labels_to_levels() allows to use value labels of factors as their levels.

\section*{Value}
\(x\), where for all factors former levels are replaced by their value labels.

\section*{Examples}
```

data(efc)

# create factor

x <- as.factor(efc\$c172code)

# add value labels - these are not factor levels yet

x <- assign_labels(x, values = c(`1` = "low", `2` = "mid", `3` = "high"))
levels(x)
data_tabulate(x)
x <- labels_to_levels(x)
levels(x)
data_tabulate(x)

```
```

makepredictcall.dw_transformer
Utility Function for Safe Prediction with datawizard transformers

```

\section*{Description}

This function allows for the use of (some of) datawizard's transformers inside a model formula. See examples below.

Currently, center(), standardize(), normalize(), \& rescale() are supported.

\section*{Usage}
\#\# S3 method for class 'dw_transformer'
makepredictcall(var, call)

\section*{Arguments}
var A variable.
call The term in the formula, as a call.

\section*{Value}

A replacement for call for the predvars attribute of the terms.

\section*{See Also}
```

stats::makepredictcall()

```

\section*{Examples}
```

data("mtcars")
train <- mtcars[1:30, ]
test <- mtcars[31:32, ]
m1 <- lm(mpg ~ center(hp), data = train)
predict(m1, newdata = test) \# Data is "centered" before the prediction is made,

# according to the center of the old data

m2 <- lm(mpg ~ standardize(hp), data = train)
m3 <- lm(mpg ~ scale(hp), data = train) \# same as above
predict(m2, newdata = test) \# Data is "standardized" before the prediction is made.
predict(m3, newdata = test) \# Data is "standardized" before the prediction is made.
m4 <- lm(mpg ~ normalize(hp), data = mtcars)
m5 <- lm(mpg ~ rescale(hp, to = c(-3, 3)), data = mtcars)
(newdata <- data.frame(hp = c(range(mtcars\$hp), 400))) \# 400 is outside original range!
model.frame(delete.response(terms(m4)), data = newdata)
model.frame(delete.response(terms(m5)), data = newdata)

```
```

means_by_group Summary of mean values by group

```

\section*{Description}

Computes summary table of means by groups.

\section*{Usage}
means_by_group (x, ...)
\#\# S3 method for class 'numeric'
means_by_group(
x ,
by = NULL,
ci \(=0.95\),
weights \(=\) NULL,
```

    digits = NULL,
    group = NULL,
    )

## S3 method for class 'data.frame'

means_by_group(
x,
select = NULL,
by = NULL,
ci = 0.95,
weights = NULL,
digits = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
group = NULL,
)

```

\section*{Arguments}
x
... Currently not used
by If x is a numeric vector, by should be a factor that indicates the group-classifying categories. If \(x\) is a data frame, by should be a character string, naming the variable in x that is used for grouping. Numeric vectors are coerced to factors. Not that by should only refer to a single variable.
ci Level of confidence interval for mean estimates. Default is 0.95 . Use ci \(=\) NA to suppress confidence intervals.
weights If x is a numeric vector, weights should be a vector of weights that will be applied to weight all observations. If x is a data frame, weights can also be a character string indicating the name of the variable in x that should be used for weighting. Default is NULL, so no weights are used.
digits Optional scalar, indicating the amount of digits after decimal point when rounding estimates and values.
group Deprecated. Use by instead.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex(""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function \((x)\) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex \(=\) TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select \(=\) regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.

\section*{Details}

This function is comparable to aggregate( x , by, mean), but provides some further information, including summary statistics from a One-Way-ANOVA using \(x\) as dependent and by as independent variable. emmeans: :contrast() is used to get p -values for each sub-group. P-values indicate whether each group-mean is significantly different from the total mean.

\section*{Value}

A data frame with information on mean and further summary statistics for each sub-group.

\section*{Examples}
```

data(efc)

```
```

means_by_group(efc, "c12hour", "e42dep")
data(iris)
means_by_group(iris, "Sepal.Width", "Species")

# weighting

efc\$weight <- abs(rnorm(n = nrow(efc), mean = 1, sd = .5))
means_by_group(efc, "c12hour", "e42dep", weights = "weight")

```
mean_sd Summary Helpers

\section*{Description}

\section*{Summary Helpers}

\section*{Usage}
```

mean_sd(x, times = 1L, remove_na = TRUE, named = TRUE, na.rm = TRUE, ...)
median_mad(
x ,
times = 1L,
remove_na = TRUE,
constant $=1.4826$,
named = TRUE,
na.rm = TRUE,
)

```

\section*{Arguments}
x
times
remove_na
constant
named \(\quad\) Should the vector be named? (E.g., c \(("-S D "=-1\), Mean = 1, " + SD" = 2).)
na.rm Deprecated. Please use remove_na instead. Not used.
A numeric vector (or one that can be coerced to one via as. numeric()) to be summarized.

How many SDs above and below the Mean (or MADs around the Median)
Logical. Should NA values be removed before computing (TRUE) or not (FALSE, default)?
scale factor.

\section*{Value}

A (possibly named) numeric vector of length \(2 *\) times +1 of SDs below the mean, the mean, and SDs above the mean (or median and MAD).

\section*{Examples}
```

mean_sd(mtcars$mpg)
mean_sd(mtcars$mpg, times = 2L)
median_mad(mtcars\$mpg)

```
nhanes_sample Sample dataset from the National Health and Nutrition Examination Survey

\section*{Description}

Selected variables from the National Health and Nutrition Examination Survey that are used in the example from Lumley (2010), Appendix E.

\section*{References}

Lumley T (2010). Complex Surveys: a guide to analysis using R. Wiley
```

normalize Normalize numeric variable to 0-1 range

```

\section*{Description}

Performs a normalization of data, i.e., it scales variables in the range \(0-1\). This is a special case of rescale(). unnormalize() is the counterpart, but only works for variables that have been normalized with normalize().

\section*{Usage}
```

normalize(x, ...)

## S3 method for class 'numeric'

normalize(x, include_bounds = TRUE, verbose = TRUE, ...)
\#\# S3 method for class 'data.frame'
normalize(
x,
select = NULL,
exclude = NULL,
include_bounds = TRUE,
append = FALSE,
ignore_case = FALSE,

```
```

    regex = FALSE,
    verbose = TRUE,
    )
unnormalize(x, ...)

## S3 method for class 'numeric'

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

## S3 method for class 'data.frame'

unnormalize(
x,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
...
)

## S3 method for class 'grouped_df'

unnormalize(
x,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}
x
... Arguments passed to or from other methods.
include_bounds Numeric or logical. Using this can be useful in case of beta-regression, where the response variable is not allowed to include zeros and ones. If TRUE, the input is normalized to a range that includes zero and one. If FALSE, the return value is compressed, using Smithson and Verkuilen's (2006) formula ( \(x\) * \((n-1)+0.5) / n\), to avoid zeros and ones in the normalized variables. Else, if numeric (e.g., 0.001), include_bounds defines the "distance" to the lower and upper bound, i.e. the normalized vectors are rescaled to a range from \(0+\) include_bounds to 1 - include_bounds.
verbose Toggle warnings and messages on or off.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
append Logical or string. If TRUE, standardized variables get new column names (with the suffix "_z") and are appended (column bind) to \(x\), thus returning both the original and the standardized variables. If FALSE, original variables in \(x\) will be overwritten by their standardized versions. If a character value, standardized variables are appended with new column names (using the defined suffix) to the original data frame.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Details}
- If x is a matrix, normalization is performed across all values (not column- or row-wise). For column-wise normalization, convert the matrix to a data.frame.
- If \(x\) is a grouped data frame (grouped_df), normalization is performed separately for each group.

\section*{Value}

A normalized object.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{References}

Smithson M, Verkuilen J (2006). A Better Lemon Squeezer? Maximum-Likelihood Regression with Beta-Distributed Dependent Variables. Psychological Methods, 11(1), 54-71.

\section*{See Also}

See makepredictcall.dw_transformer() for use in model formulas.
Other transform utilities: ranktransform(), rescale(), reverse(), standardize()

\section*{Examples}
```

normalize(c(0, 1, 5, -5, -2))
normalize(c(0, 1, 5, -5, -2), include_bounds = FALSE)

# use a value defining the bounds

normalize(c(0, 1, 5, -5, -2), include_bounds = .001)
head(normalize(trees))

```

\section*{Description}

Transform numeric values with the integers of their rank (i.e., 1st smallest, 2nd smallest, 3rd smallest, etc.). Setting the sign argument to TRUE will give you signed ranks, where the ranking is done according to absolute size but where the sign is preserved (i.e., \(2,1,-3,4\) ).

\section*{Usage}
```

ranktransform(x, ...)
\#\# S3 method for class 'numeric'
ranktransform(x, sign = FALSE, method = "average", verbose = TRUE, ...)
\#\# S3 method for class 'data.frame'
ranktransform(
x ,
select $=$ NULL,
exclude = NULL,
sign = FALSE,
method = "average",
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}

X
...
sign
method
verbose
select

Object.
Arguments passed to or from other methods.
Logical, if TRUE, return signed ranks.
Treatment of ties. Can be one of "average" (default), "first", "last", "random", "max" or "min". See rank() for details.
Toggle warnings.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is. numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or
-(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Value}

A rank-transformed object.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{See Also}

Other transform utilities: normalize(), rescale(), reverse(), standardize()

\section*{Examples}
```

ranktransform(c(0, 1, 5, -5, -2))

# Won't work

# ranktransform(c(0, 1, 5, -5, -2), sign = TRUE)

head(ranktransform(trees))

```
```

recode_into

```

\section*{Description}

This functions recodes values from one or more variables into a new variable. It is a convenient function to avoid nested ifelse() statements, which is similar to dplyr::case_when().

\section*{Usage}
```

recode_into(
...,
data = NULL,
default = NA,
overwrite = TRUE,
preserve_na = FALSE,
verbose = TRUE
)

```

\section*{Arguments}
\begin{tabular}{ll}
\(\ldots\). & \begin{tabular}{l} 
A sequence of two-sided formulas, where the left hand side (LHS) is a logical \\
matching condition that determines which values match this case. The LHS of \\
this formula is also called "recode pattern" (e.g., in messages). The right hand \\
side (RHS) indicates the replacement value.
\end{tabular} \\
data & \begin{tabular}{l} 
Optional, name of a data frame. This can be used to avoid writing the data name \\
multiple times in . . . See 'Examples'. \\
default \\
Indicates the default value that is chosen when no match in the formulas in . . . \\
is found. If not provided, NA is used as default value.
\end{tabular} \\
overwrite & \begin{tabular}{l} 
Logical, if TRUE (default) and more than one recode pattern apply to the same \\
case, already recoded values will be overwritten by subsequent recode patterns.
\end{tabular} \\
If FALSE, former recoded cases will not be altered by later recode patterns that \\
would apply to those cases again. A warning message is printed to alert such \\
preserve_na & \begin{tabular}{l} 
Lituations and to avoid unintentional recodings. \\
Logical, if TRUE and default is not NA, missing values in the original variable \\
will be set back to NA in the recoded variable (unless overwritten by other recode \\
patterns). If FALSE, missing values in the original variable will be recoded to \\
default. Setting preserve_na = TRUE prevents unintentional overwriting of
\end{tabular} \\
missing values with default, which means that you won't find valid values
\end{tabular}

\section*{Value}

A vector with recoded values.

\section*{Examples}
```

x <- 1:30
recode_into(
x > 15 ~ "a",
x > 10 \& x <= 15 ~ "b",
default = "c"
)
x <- 1:10

# default behaviour: second recode pattern "x > 5" overwrites

# some of the formerly recoded cases from pattern "x >= 3 \& x <= 7"

recode_into(
x >= 3 \& x <= 7 ~ 1,
x > 5 ~ 2,
default = 0,
verbose = FALSE
)

# setting "overwrite = FALSE" will not alter formerly recoded cases

recode_into(
x >= 3 \& x <= 7 ~ 1,
x > 5 ~ 2,
default = 0,
overwrite = FALSE,
verbose = FALSE
)
set.seed(123)
d <- data.frame(
x = sample(1:5, 30, TRUE),
y = sample(letters[1:5], 30, TRUE),
stringsAsFactors = FALSE
)

# from different variables into new vector

recode_into(
d$x %in% 1:3 & d$y %in% c("a", "b") ~ 1,
d\$x > 3 ~ 2,
default = 0
)

# no need to write name of data frame each time

recode_into(
x %in% 1:3 \& y %in% c("a", "b") ~ 1,
x > 3 ~ 2,
data = d,
default = 0
)

# handling of missing values

d <- data.frame(
x = c(1, NA, 2, NA, 3, 4),

```
recode_values
```

    y = c(1, 11, 3, NA, 5, 6)
    )

# first NA in x is overwritten by valid value from y

# we have no known value for second NA in x and y,

# thus we get one NA in the result

recode_into(
x <= 3 ~ 1,
y > 5 ~ 2,
data = d,
default = 0,
preserve_na = TRUE
)

# first NA in x is overwritten by valid value from y

# default value is used for second NA

recode_into(
x <= 3 ~ 1,
y > 5 ~ 2,
data = d,
default = 0,
preserve_na = FALSE
)

```
```

recode_values Recode old values of variables into new values

```

\section*{Description}

This functions recodes old values into new values and can be used to to recode numeric or character vectors, or factors.

\section*{Usage}
```

recode_values(x, ...)

## S3 method for class 'numeric'

recode_values(
x,
recode = NULL,
default = NULL,
preserve_na = TRUE,
verbose = TRUE,
)
\#\# S3 method for class 'data.frame'
recode_values(
x,
select = NULL,
exclude = NULL,

```
```

    recode = NULL,
    default = NULL,
    preserve_na = TRUE,
    append = FALSE,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
    ...
    )
change_code(
x,
select = NULL,
exclude = NULL,
recode = NULL,
default = NULL,
preserve_na = TRUE,
append = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
...
)

```

\section*{Arguments}
x
...
recode
default Defines the default value for all values that have no match in the recode-pairs. Note that, if preserve_na=FALSE, missing values (NA) are also captured by the default argument, and thus will also be recoded into the specified value. See 'Examples' and 'Details'.
preserve_na Logical, if TRUE, NA (missing values) are preserved. This overrides any other arguments, including default. Hence, if preserve_na=TRUE, default will no longer convert NA into the specified default value.
verbose Toggle warnings.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
append Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in \(x\) will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Details}

This section describes the pattern of the recode arguments, which also provides some shortcuts, in particular when recoding numeric values.
- Single values

Single values either need to be wrapped in backticks (in case of numeric values) or "as is" (for character or factor levels). Example: recode=list ( \(0^{-}=1,{ }^{`}{ }^{`}=2\) ) would recode 1 into 0 , and 2 into 1 . For factors or character vectors, an example is: recode=list ( \(x=" a ", y=" b "\) ) (recode "a" into "x" and "b" into "y").
- Multiple values

Multiple values that should be recoded into a new value can be separated with comma. Example: recode=list \(\left(` 1^{`}=c(1,4), 2^{`}=c(2,3)\right)\) would recode the values 1 and 4 into 1 , and 2 and 3 into 2 . It is also possible to define the old values as a character string, like: recode=list ( \(\left.1^{\prime}=" 1,4^{\prime \prime}, 2^{`}=" 2,3^{\prime \prime}\right)\) For factors or character vectors, an example is: recode=list ( \(x=c(" a ", " b ")\),
- Value range

Numeric value ranges can be defined using the :. Example: recode=list (`1`=1:3, `2`=4:6) would recode all values from 1 to 3 into 1 , and 4 to 6 into 2 .
- min and max
placeholder to use the minimum or maximum value of the (numeric) variable. Useful, e.g., when recoding ranges of values. Example: recode=list(`1`="min:10", \(2 `=" 11: m a x ")\).
- default values

The default argument defines the default value for all values that have no match in the recodepairs. For example, recode=list( \(\left.{ }^{-} 1^{`}=c(1,2), 2^{`}=c(3,4)\right)\), default=9 would recode values 1 and 2 into 1,3 and 4 into 2 , and all other values into 9 . If preserve_na is set to FALSE, NA (missing values) will also be recoded into the specified default value.
- Reversing and rescaling

See reverse() and rescale().

\section*{Value}
x , where old values are replaced by new values.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{Note}

You can use options(data_recode_pattern = "old=new") to switch the behaviour of the recodeargument, i.e. recode-pairs are now following the pattern old values \(=\) new values, e.g. if getOption("data_recode_pattern") is set to "old=new", then recode(`1`=0) would recode all 1 into 0 . The default for recode \(\left({ }^{-} 1^{`}=0\right)\) is to recode all 0 into 1 .

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```


# numeric ----------

set.seed(123)
x <- sample(c(1:4,NA), 15, TRUE)
table(x, useNA = "always")
out <- recode_values(x, list(`0` = 1, `1` = 2:3, `2` = 4))
out
table(out, useNA = "always")

# to recode NA values, set preserve_na to FALSE

out <- recode_values(
x,
list(`0` = 1, `1` = 2:3, `2` = 4, `9` = NA),
preserve_na = FALSE
)
out
table(out, useNA = "always")

# preserve na ----------

out <- recode_values(x, list(`0` = 1, `1` = 2:3), default = 77)
out
table(out, useNA = "always")

# recode na into default ----------

out <- recode_values(
x,
list(`0` = 1, `1` = 2:3),
default = 77,
preserve_na = FALSE
)
out
table(out, useNA = "always")

# factors (character vectors are similar) ----------

set.seed(123)
x <- as.factor(sample(c("a", "b", "c"), 15, TRUE))
table(x)
out <- recode_values(x, list(x = "a", y = c("b", "c")))
out

```
```

table(out)
out <- recode_values(x, list(x = "a", y = "b", z = "c"))
out
table(out)
out <- recode_values(x, list(y = "b,c"), default = 77)

# same as

# recode_values(x, list(y = c("b", "c")), default = 77)

out
table(out)

# data frames ----------

set.seed(123)
d <- data.frame(
x = sample(c(1:4, NA), 12, TRUE),
y = as.factor(sample(c("a", "b", "c"), 12, TRUE)),
stringsAsFactors = FALSE
)
recode_values(
d,
recode = list(`0` = 1, `1` = 2:3, `2` = 4, x = "a", y = c("b", "c")),
append = TRUE
)

# switch recode pattern to "old=new" ----------

options(data_recode_pattern = "old=new")

# numeric

set.seed(123)
x <- sample(c(1:4,NA), 15, TRUE)
table(x, useNA = "always")
out <- recode_values(x, list(`1` = 0, `2:3` = 1, `4` = 2))
table(out, useNA = "always")

# factors (character vectors are similar)

set.seed(123)
x <- as.factor(sample(c("a", "b", "c"), 15, TRUE))
table(x)
out <- recode_values(x, list(a = "x", `b, c` = "y"))
table(out)

# reset options

options(data_recode_pattern = NULL)

```
```

remove_empty Return or remove variables or observations that are completely miss-
ing

```

\section*{Description}

These functions check which rows or columns of a data frame completely contain missing values, i.e. which observations or variables completely have missing values, and either (1) returns their indices; or (2) removes them from the data frame.

\section*{Usage}
empty_columns(x)
empty_rows(x)
remove_empty_columns(x)
remove_empty_rows(x)
remove_empty (x)

\section*{Arguments}
x
A data frame.

\section*{Details}

For character vectors, empty string values (i.e. "") are also considered as missing value. Thus, if a character vector only contains NA and "" \({ }^{-\cdots}\), it is considered as empty variable and will be removed. Same applie

\section*{Value}
- For empty_columns() and empty_rows(), a numeric (named) vector with row or column indices of those variables that completely have missing values.
- For remove_empty_columns() and remove_empty_rows(), a data frame with "empty" columns or rows removed, respectively.
- For remove_empty (), both empty rows and columns will be removed.

\section*{Examples}
```

tmp <- data.frame(
a = c(1, 2, 3, NA, 5),
b = c(1, NA, 3, NA, 5),
c = c(NA,NA,NA,NA,NA),
d = c(1,NA, 3,NA, 5)
)
tmp

```
```


# indices of empty columns or rows

empty_columns(tmp)
empty_rows(tmp)

# remove empty columns or rows

remove_empty_columns(tmp)
remove_empty_rows(tmp)

# remove empty columns and rows

remove_empty(tmp)

# also remove "empty" character vectors

tmp <- data.frame(
a = c(1, 2, 3, NA, 5),
b = c(1, NA, 3, NA, 5),
c = c("", "", "", "", ""),
stringsAsFactors = FALSE
)
empty_columns(tmp)

```
replace_nan_inf

Convert infinite or NaN values into NA

\section*{Description}

Replaces all infinite (Inf and -Inf) or NaN values with NA.

\section*{Usage}
replace_nan_inf(x, ...)

\section*{Arguments}
\begin{tabular}{ll}
\(x\) & A vector or a dataframe \\
\(\ldots\) & Currently not used.
\end{tabular}

\section*{Value}

Data with Inf, -Inf, and NaN converted to NA.

\section*{Examples}
```


# a vector

x <- c(1, 2, NA, 3, NaN, 4, NA, 5, Inf, -Inf, 6, 7)
replace_nan_inf(x)

# a data frame

```
rescale
```

    df <- data.frame(
        x = c(1, NA, 5, Inf, 2, NA),
    y = c(3, NaN, 4, -Inf, 6, 7),
    stringsAsFactors = FALSE
    )
replace_nan_inf(df)

```
rescale Rescale Variables to a New Range

\section*{Description}

Rescale variables to a new range. Can also be used to reverse-score variables (change the keying/scoring direction), or to expand a range.

\section*{Usage}
```

rescale(x, ...)
change_scale(x, ...)
\#\# S3 method for class 'numeric'
rescale(
x,
to $=c(0,100)$,
multiply = NULL,
add $=$ NULL,
range $=$ NULL,
verbose = TRUE,
)
\#\# S3 method for class 'data.frame'
rescale(
x ,
select $=$ NULL,
exclude $=$ NULL,
to $=c(0,100)$,
multiply = NULL,
add $=$ NULL,
range $=$ NULL,
append = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = FALSE,
)

```

\section*{Arguments}
multiply If not NULL, to is ignored and multiply will be used, giving the factor by which

X
.
to
add
range
verbose
select

A (grouped) data frame, numeric vector or factor.
Arguments passed to or from other methods.
Numeric vector of length 2 giving the new range that the variable will have after rescaling. To reverse-score a variable, the range should be given with the maximum value first. See examples. the actual range of \(x\) should be expanded. For example, if a vector ranges from 5 to 15 and multiply \(=1.1\), the current range of 10 will be expanded by the factor of 1.1 , giving a new range of 11 . Thus, the rescaled vector would range from 4.5 to 15.5 .

A vector of length 1 or 2 . If not NULL, to is ignored and add will be used, giving the amount by which the minimum and maximum of the actual range of \(x\) should be expanded. For example, if a vector ranges from 5 to 15 and add \(=1\), the range will be expanded from 4 to 16 . If add is of length 2 , then the first value is used for the lower bound and the second value for the upper bound.

Initial (old) range of values. If NULL, will take the range of the input vector (range (x)).

Toggle warnings.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
\begin{tabular}{|c|c|}
\hline & If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species". \\
\hline exclude & See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns. \\
\hline append & Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in \(x\) will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame. \\
\hline ignore_case & Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names. \\
\hline regex & Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex \(=\) TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex(""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround. \\
\hline
\end{tabular}

\section*{Value}

A rescaled object.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{See Also}

See makepredictcall.dw_transformer() for use in model formulas.
Other transform utilities: normalize(), ranktransform(), reverse(), standardize()

\section*{Examples}
```

rescale(c(0, 1, 5, -5, -2))
rescale(c(0, 1, 5, -5, -2), to = c(-5, 5))
rescale(c(1, 2, 3, 4, 5), to = c(-2, 2))

# Specify the "theoretical" range of the input vector

rescale(c(1, 3, 4), to =c(0, 40), range = c(0, 4))

```
```


# Reverse-score a variable

rescale(c(1, 2, 3, 4, 5), to = c(5, 1))
rescale(c(1, 2, 3, 4, 5), to = c(2, -2))

# Data frames

head(rescale(iris, to = c(0, 1)))
head(rescale(iris, to = c(0, 1), select = "Sepal.Length"))

# One can specify a list of ranges

head(rescale(iris, to = list(
"Sepal.Length" = c(0, 1),
"Petal.Length" = c(-1, 0)
)))

# "expand" ranges by a factor or a given value

x <- 5:15
x

# both will expand the range by 10%

rescale(x, multiply = 1.1)
rescale(x, add = 0.5)

# expand range by different values

rescale(x, add = c(1, 3))

# Specify list of multipliers

d <- data.frame(x = 5:15, y = 5:15)
rescale(d, multiply = list(x = 1.1, y = 0.5))

```
rescale_weights

\section*{Description}

Most functions to fit multilevel and mixed effects models only allow to specify frequency weights, but not design (i.e. sampling or probability) weights, which should be used when analyzing complex samples and survey data. rescale_weights() implements an algorithm proposed by Asparouhov (2006) and Carle (2009) to rescale design weights in survey data to account for the grouping structure of multilevel models, which then can be used for multilevel modelling.

\section*{Usage}
rescale_weights(data, by, probability_weights, nest = FALSE, group = NULL)

\section*{Arguments}
\[
\text { data } \quad \text { A data frame. }
\]
by Variable names (as character vector, or as formula), indicating the grouping structure (strata) of the survey data (level-2-cluster variable). It is also possible to create weights for multiple group variables; in such cases, each created weighting variable will be suffixed by the name of the group variable.
probability_weights
Variable indicating the probability (design or sampling) weights of the survey data (level-1-weight).
nest Logical, if TRUE and by indicates at least two group variables, then groups are "nested", i.e. groups are now a combination from each group level of the variables in by.
group Deprecated. Use by instead.

\section*{Details}

Rescaling is based on two methods: For pweights_a, the sample weights probability_weights are adjusted by a factor that represents the proportion of group size divided by the sum of sampling weights within each group. The adjustment factor for pweights_b is the sum of sample weights within each group divided by the sum of squared sample weights within each group (see Carle (2009), Appendix B). In other words, pweights_a "scales the weights so that the new weights sum to the cluster sample size" while pweights_b "scales the weights so that the new weights sum to the effective cluster size".

Regarding the choice between scaling methods A and B, Carle suggests that "analysts who wish to discuss point estimates should report results based on weighting method A. For analysts more interested in residual between-group variance, method B may generally provide the least biased estimates". In general, it is recommended to fit a non-weighted model and weighted models with both scaling methods and when comparing the models, see whether the "inferential decisions converge", to gain confidence in the results.

Though the bias of scaled weights decreases with increasing group size, method A is preferred when insufficient or low group size is a concern.

The group ID and probably PSU may be used as random effects (e.g. nested design, or group and PSU as varying intercepts), depending on the survey design that should be mimicked.

\section*{Value}
data, including the new weighting variables: pweights_a and pweights_b, which represent the rescaled design weights to use in multilevel models (use these variables for the weights argument).

\section*{References}
- Carle A.C. (2009). Fitting multilevel models in complex survey data with design weights: Recommendations. BMC Medical Research Methodology 9(49): 1-13
- Asparouhov T. (2006). General Multi-Level Modeling with Sampling Weights. Communications in Statistics - Theory and Methods 35: 439-460

\section*{Examples}
```

if (require("lme4")) {
data(nhanes_sample)
head(rescale_weights(nhanes_sample, "SDMVSTRA", "WTINT2YR"))
\# also works with multiple group-variables
head(rescale_weights(nhanes_sample, c("SDMVSTRA", "SDMVPSU"), "WTINT2YR"))
\# or nested structures.
x <- rescale_weights(
data = nhanes_sample,
by = c("SDMVSTRA", "SDMVPSU"),
probability_weights = "WTINT2YR",
nest = TRUE
)
head(x)
nhanes_sample <- rescale_weights(nhanes_sample, "SDMVSTRA", "WTINT2YR")
glmer(
total ~ factor(RIAGENDR) * (log(age) + factor(RIDRETH1)) + (1 | SDMVPSU),
family = poisson(),
data = nhanes_sample,
weights = pweights_a
)
}

```
reshape_ci Reshape CI between wide/long formats

\section*{Description}

Reshape CI between wide/long formats.

\section*{Usage}
reshape_ci(x, ci_type = "CI")

\section*{Arguments}
\(x \quad\) A data frame containing columns named CI_low and CI_high (or similar, see ci_type).
ci_type String indicating the "type" (i.e. prefix) of the interval columns. Per easystats convention, confidence or credible intervals are named CI_low and CI_high, and the related ci_type would be "CI". If column names for other intervals differ, ci_type can be used to indicate the name, e.g. ci_type = "SI" can be used for support intervals, where the column names in the data frame would be SI_low and SI_high.

\section*{Value}

A data frame with columns corresponding to confidence intervals reshaped either to wide or long format.

\section*{Examples}
```

x <- data.frame(
Parameter = c("Term 1", "Term 2", "Term 1", "Term 2"),
CI = c(.8, .8, .9, .9),
CI_low = c(.2, .3, .1, .15),
CI_high = c(.5, .6, .8, .85),
stringsAsFactors = FALSE
)
reshape_ci(x)
reshape_ci(reshape_ci(x))

```
    reverse Reverse-Score Variables

\section*{Description}

Reverse-score variables (change the keying/scoring direction).

\section*{Usage}
```

reverse(x, ...)
reverse_scale(x, ...)
\#\# S3 method for class 'numeric'
reverse (x, range $=$ NULL, verbose $=$ TRUE, ...)
\#\# S3 method for class 'data.frame'
reverse(
x ,
select $=$ NULL,
exclude = NULL,
range $=$ NULL,
append = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = FALSE,
)

```

\section*{Arguments}
exclude See select, however, column names matched by the pattern from exclude will

X
range
verbose
select
append

A (grouped) data frame, numeric vector or factor.
Arguments passed to or from other methods.
Range of values that is used as reference for reversing the scale. For numeric variables, can be NULL or a numeric vector of length two, indicating the lowest and highest value of the reference range. If NULL, will take the range of the input vector (range(x)). For factors, range can be NULL, a numeric vector of length two, or a (numeric) vector of at least the same length as factor levels (i.e. must be equal to or larger than nlevels(x)). Note that providing a range for factors usually only makes sense when factor levels are numeric, not characters.
Toggle warnings.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or c(1, 3, 5)),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex(""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species". be excluded instead of selected. If NULL (the default), excludes no columns.

Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or
"_s" for slide(). If append=FALSE, original variables in x will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Value}

A reverse-scored object.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{See Also}

Other transform utilities: normalize(), ranktransform(), rescale(), standardize()

\section*{Examples}
```

reverse(c(1, 2, 3, 4, 5))
reverse(c(-2, -1, 0, 2, 1))

# Specify the "theoretical" range of the input vector

reverse(c(1, 3, 4), range = c(0, 4))

# Factor variables

reverse(factor(c(1, 2, 3, 4, 5)))
reverse(factor(c(1, 2, 3, 4, 5)), range = 0:10)

# Data frames

head(reverse(iris))
head(reverse(iris, select = "Sepal.Length"))

```

\section*{Description}

Tools for working with row names or row ids

\section*{Usage}
rownames_as_column(x, var = "rowname")
column_as_rownames(x, var = "rowname")
rowid_as_column(x, var = "rowid")

\section*{Arguments}
x
var Name of column to use for row names/ids. For column_as_rownames(), this argument can be the variable name or the column number. For rownames_as_column() and rowid_as_column(), the column name must not already exist in the data.

\section*{Details}

These are similar to tibble's functions column_to_rownames(), rownames_to_column() and rowid_to_column(). Note that the behavior of rowid_as_column() is different for grouped dataframe: instead of making the rowid unique across the full dataframe, it creates rowid per group. Therefore, there can be several rows with the same rowid if they belong to different groups.
If you are familiar with dplyr, this is similar to doing the following:
```

data |>
group_by(grp) |>
mutate(id = row_number()) |>
ungroup()

```

\section*{Value}

A data frame.

\section*{Examples}
```


# Convert between row names and column -----------------------------------

test <- rownames_as_column(mtcars, var = "car")
test
head(column_as_rownames(test, var = "car"))
test_data <- head(iris)

```
```

rowid_as_column(test_data)
rowid_as_column(test_data, var = "my_id")

```
row_means Row means (optionally with minimum amount of valid values)

\section*{Description}

This function is similar to the SPSS MEAN. n function and computes row means from a data frame or matrix if at least min_valid values of a row are valid (and not NA).

\section*{Usage}
```

row_means(
data,
select = NULL,
exclude = NULL,
min_valid = NULL,
digits = NULL,
ignore_case = FALSE,
regex = FALSE,
remove_na = FALSE,
verbose = TRUE
)

```

\section*{Arguments}

\section*{data}

A data frame with at least two columns, where row means are applied.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., \(\sim\) column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex ()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.

If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
min_valid Optional, a numeric value of length 1. May either be
- a numeric value that indicates the amount of valid values per row to calculate the row mean;
- or a value between 0 and 1 , indicating a proportion of valid values per row to calculate the row mean (see 'Details').
- NULL (default), in which all cases are considered.

If a row's sum of valid values is less than min_valid, NA will be returned.
digits \(\quad\) Numeric value indicating the number of decimal places to be used for rounding mean values. Negative values are allowed (see 'Details'). By default, digits = NULL and no rounding is used.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
remove_na Logical, if TRUE (default), removes missing (NA) values before calculating row means. Only applies if min_valuid is not specified.
verbose Toggle warnings.

\section*{Details}

Rounding to a negative number of digits means rounding to a power of ten, for example row_means (df, 3 , digits \(=-2\) ) rounds to the nearest hundred. For min_valid, if not NULL, min_valid must be a numeric value from 0 to ncol (data). If a row in the data frame has at least min_valid non-missing values, the row mean is returned. If min_valid is a non-integer value from 0 to 1 , min_valid is considered to indicate the proportion of required non-missing values per row. E.g., if min_valid = 0.75 , a row must have at least ncol (data) * min_valid non-missing values for the row mean to be calculated. See 'Examples'.

\section*{Value}

A vector with row means for those rows with at least \(n\) valid values.

\section*{Examples}
```

dat <- data.frame(
c1 = c(1, 2, NA, 4),
c2 = c(NA, 2, NA, 5),
c3 = c(NA, 4, NA, NA),
c4 = c(2, 3, 7, 8)
)

# default, all means are shown, if no NA values are present

row_means(dat)

# remove all NA before computing row means

row_means(dat, remove_na = TRUE)

# needs at least 4 non-missing values per row

row_means(dat, min_valid = 4) \# 1 valid return value

# needs at least 3 non-missing values per row

row_means(dat, min_valid = 3) \# 2 valid return values

# needs at least 2 non-missing values per row

row_means(dat, min_valid = 2)

# needs at least 1 non-missing value per row, for two selected variables

row_means(dat, select = c("c1", "c3"), min_valid = 1)

# needs at least 50% of non-missing values per row

row_means(dat, min_valid = 0.5) \# 3 valid return values

# needs at least 75% of non-missing values per row

row_means(dat, min_valid = 0.75) \# 2 valid return values

```
row_to_colnames Tools for working with column names

\section*{Description}

Tools for working with column names

\section*{Usage}
row_to_colnames(x, row = 1, na_prefix = "x", verbose = TRUE)
colnames_to_row(x, prefix = "x")

\section*{Arguments}

X
row Row to use as column names.
na_prefix Prefix to give to the column name if the row has an NA. Default is ' \(x\) ', and it will be incremented at each NA ( \(x 1, x 2\), etc.).
verbose Toggle warnings.
prefix Prefix to give to the column name. Default is ' \(x\) ', and it will be incremented at each column ( \(\mathrm{x} 1, \mathrm{x} 2\), etc.).

\section*{Value}
row_to_colnames() and colnames_to_row() both return a data frame.

\section*{Examples}
```


# Convert a row to column names -----------------------------------

test <- data.frame(
a = c("iso", 2, 5),
b = c("year", 3, 6),
c = c("value", 5, 7)
)
test
row_to_colnames(test)

# Convert column names to row -----------------------------------

test <- data.frame(
ARG = c("BRA", "FRA"),
``1960`= c(1960, 1960),        `2000` = c(2000, 2000)
)
test
colnames_to_row(test)

```
skewness

Compute Skewness and (Excess) Kurtosis

\section*{Description}

Compute Skewness and (Excess) Kurtosis

\section*{Usage}
skewness(x, ...)
\#\# S3 method for class 'numeric'
skewness(
```

    x,
    remove_na = TRUE,
    type = "2",
    iterations = NULL,
    verbose = TRUE,
    na.rm = TRUE,
    ...
    )
kurtosis(x, ...)

## S3 method for class 'numeric'

kurtosis(
x,
remove_na = TRUE,
type = "2",
iterations = NULL,
verbose = TRUE,
na.rm = TRUE,
)

## S3 method for class 'parameters_kurtosis'

print(x, digits = 3, test = FALSE, ...)

## S3 method for class 'parameters_skewness'

print(x, digits = 3, test = FALSE, ...)

## S3 method for class 'parameters_skewness'

summary(object, test = FALSE, ...)

## S3 method for class 'parameters_kurtosis'

summary(object, test = FALSE, ...)

```

\section*{Arguments}
\begin{tabular}{ll}
\(x\) & A numeric vector or data.frame. \\
\(\ldots\) & Arguments passed to or from other methods. \\
remove_na & \begin{tabular}{l} 
Logical. Should NA values be removed before computing (TRUE) or not (FALSE, \\
default)?
\end{tabular} \\
type & \begin{tabular}{l} 
Type of algorithm for computing skewness. May be one of 1 (or " "1", "I" \\
or "classic"), 2 (or "2", "II" or "SPSS" or "SAS") or 3 (or " " \(3 ", ~ " I I I " ~ o r ~\)
\end{tabular} \\
"Minitab"). See 'Details'.
\end{tabular}
\begin{tabular}{ll} 
digits & Number of decimal places. \\
test & Logical, if TRUE, tests if skewness or kurtosis is significantly different from zero. \\
object & An object returned by skewness() or kurtosis().
\end{tabular}

\section*{Details}

Skewness: Symmetric distributions have a skewness around zero, while a negative skewness values indicates a "left-skewed" distribution, and a positive skewness values indicates a "rightskewed" distribution. Examples for the relationship of skewness and distributions are:
- Normal distribution (and other symmetric distribution) has a skewness of 0
- Half-normal distribution has a skewness just below 1
- Exponential distribution has a skewness of 2
- Lognormal distribution can have a skewness of any positive value, depending on its parameters
(https://en.wikipedia.org/wiki/Skewness)

Types of Skewness: skewness() supports three different methods for estimating skewness, as discussed in Joanes and Gill (1988):
- Type " 1 " is the "classical" method, which is \(g 1=\left(\operatorname{sum}\left((x-\operatorname{mean}(x))^{\wedge} 3\right) / n\right) /(\operatorname{sum}((x-\) mean \(\left.\left.(x))^{\wedge} 2\right) / n\right)^{\wedge} 1.5\)
- Type " 2 " first calculates the type-1 skewness, then adjusts the result: G1 = g1 * sqrt ( n * ( n -1)) / ( \(n-2\) ). This is what SAS and SPSS usually return.
- Type " 3 " first calculates the type-1 skewness, then adjusts the result: b1 = g1 * ( ( \(1-1\) / \(n))^{\wedge} 1.5\). This is what Minitab usually returns.

Kurtosis: The kurtosis is a measure of "tailedness" of a distribution. A distribution with a kurtosis values of about zero is called "mesokurtic". A kurtosis value larger than zero indicates a "leptokurtic" distribution with fatter tails. A kurtosis value below zero indicates a "platykurtic" distribution with thinner tails (https://en.wikipedia.org/wiki/Kurtosis).

Types of Kurtosis: kurtosis() supports three different methods for estimating kurtosis, as discussed in Joanes and Gill (1988):
- Type " 1 " is the "classical" method, which is \(\mathrm{g} 2=\mathrm{n} * \operatorname{sum}\left((x-\operatorname{mean}(x))^{\wedge} 4\right) /(\operatorname{sum}((x-\) mean \(\left.\left.(x))^{\wedge} 2\right)^{\wedge} 2\right)-3\).
- Type "2" first calculates the type-1 kurtosis, then adjusts the result: \(\mathrm{G} 2=((\mathrm{n}+1) * \mathrm{~g} 2+6)\) * \((n-1) /((n-2) *(n-3))\). This is what SAS and SPSS usually return
- Type " 3 " first calculates the type-1 kurtosis, then adjusts the result: \(\mathrm{b} 2=(\mathrm{g} 2+3) *(1-1 /\) \(n)^{\wedge} 2-3\). This is what Minitab usually returns.

Standard Errors: It is recommended to compute empirical (bootstrapped) standard errors (via the iterations argument) than relying on analytic standard errors (Wright \& Herrington, 2011).

\section*{Value}

Values of skewness or kurtosis.

\section*{References}
- D. N. Joanes and C. A. Gill (1998). Comparing measures of sample skewness and kurtosis. The Statistician, 47, 183-189.
- Wright, D. B., \& Herrington, J. A. (2011). Problematic standard errors and confidence intervals for skewness and kurtosis. Behavior research methods, 43(1), 8-17.

\section*{Examples}
skewness(rnorm(1000))
kurtosis(rnorm(1000))
slide \(\quad\) Shift numeric value range

\section*{Description}

This functions shifts the value range of a numeric variable, so that the new range starts at a given value.

\section*{Usage}
```

    slide(x, ...)
    ## S3 method for class 'numeric'
    slide(x, lowest = 0, ...)
    ## S3 method for class 'data.frame'
    slide(
        x,
        select = NULL,
        exclude = NULL,
        lowest = 0,
        append = FALSE,
        ignore_case = FALSE,
        regex = FALSE,
        verbose = TRUE,
    )
    ```

\section*{Arguments}
x
...
lowest

A data frame or numeric vector.
not used.
Numeric, indicating the lowest (minimum) value when converting factors or character vectors to numeric values.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
append Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in \(x\) will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose

\section*{Value}
\(x\), where the range of numeric variables starts at a new value.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder (), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```


# numeric

head(mtcars$gear)
head(slide(mtcars$gear))
head(slide(mtcars\$gear, lowest = 10))

# data frame

sapply(slide(mtcars, lowest = 1), min)
sapply(mtcars, min)

```
smoothness

Quantify the smoothness of a vector

\section*{Description}

Quantify the smoothness of a vector

\section*{Usage}
smoothness(x, method = "cor", lag = 1, iterations = NULL, ...)

\section*{Arguments}
\begin{tabular}{ll}
x \\
method & \begin{tabular}{l} 
Numeric vector (similar to a time series). \\
Can be "diff" (the standard deviation of the standardized differences) or "cor" \\
(default, lag-one autocorrelation).
\end{tabular} \\
lag & \begin{tabular}{l} 
An integer indicating which lag to use. If less than 1, will be interpreted as \\
expressed in percentage of the length of the vector.
\end{tabular} \\
iterations & \begin{tabular}{l} 
The number of bootstrap replicates for computing standard errors. If NULL (de- \\
fault), parametric standard errors are computed.
\end{tabular} \\
\(\ldots\) & \begin{tabular}{l} 
Arguments passed to or from other methods.
\end{tabular}
\end{tabular}

\section*{Value}

Value of smoothness.

\section*{References}
https://stats.stackexchange.com/questions/24607/how-to-measure-smoothness-of-a-time-series-in-r

\section*{Examples}
\(x<-(-10: 10)^{\wedge} 3+\operatorname{rnorm}(21,0,100)\)
plot(x)
smoothness(x, method = "cor")
smoothness(x, method = "diff")
```

standardize Standardization (Z-scoring)

```

\section*{Description}

Performs a standardization of data (z-scoring), i.e., centering and scaling, so that the data is expressed in terms of standard deviation (i.e., mean \(=0, \mathrm{SD}=1\) ) or Median Absolute Deviance (median \(=0\), MAD \(=1\) ). When applied to a statistical model, this function extracts the dataset, standardizes it, and refits the model with this standardized version of the dataset. The normalize() function can also be used to scale all numeric variables within the 0-1 range.

For model standardization, see standardize. default().

\section*{Usage}
standardize(x, ...)
standardise(x, ...)
\#\# S3 method for class 'numeric'
standardize(
```

    x,
    robust = FALSE,
    two_sd = FALSE,
    weights = NULL,
    reference = NULL,
    center = NULL,
    scale = NULL,
    verbose = TRUE,
    ...
    )

## S3 method for class 'factor'

standardize(
x,
robust = FALSE,
two_sd = FALSE,
weights = NULL,
force = FALSE,
verbose = TRUE,
...
)

## S3 method for class 'data.frame'

standardize(
x,
select = NULL,
exclude = NULL,
robust = FALSE,
two_sd = FALSE,
weights = NULL,
reference = NULL,
center = NULL,
scale = NULL,
remove_na = c("none", "selected", "all"),
force = FALSE,
append = FALSE,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)
unstandardize(x, ...)
unstandardise(x, ...)

## S3 method for class 'numeric'

unstandardize(

```
```

    x,
    center = NULL,
    scale = NULL,
    reference = NULL,
    robust = FALSE,
    two_sd = FALSE,
    ...
    )

## S3 method for class 'data.frame'

unstandardize(
x,
center = NULL,
scale = NULL,
reference = NULL,
robust = FALSE,
two_sd = FALSE,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
regex = FALSE,
verbose = TRUE,
)

```

\section*{Arguments}
x
... Arguments passed to or from other methods.
robust Logical, if TRUE, centering is done by subtracting the median from the variables and dividing it by the median absolute deviation (MAD). If FALSE, variables are standardized by subtracting the mean and dividing it by the standard deviation (SD).
two_sd If TRUE, the variables are scaled by two times the deviation (SD or MAD depending on robust). This method can be useful to obtain model coefficients of continuous parameters comparable to coefficients related to binary predictors, when applied to the predictors (not the outcome) (Gelman, 2008).
weights Can be NULL (for no weighting), or:
- For model: if TRUE (default), a weighted-standardization is carried out.
- For data.frames: a numeric vector of weights, or a character of the name of a column in the data. frame that contains the weights.
- For numeric vectors: a numeric vector of weights.
reference A data frame or variable from which the centrality and deviation will be computed instead of from the input variable. Useful for standardizing a subset or new data according to another data frame.
center, scale
- For standardize():

Numeric values, which can be used as alternative to reference to define a reference centrality and deviation. If scale and center are of length 1 , they will be recycled to match the length of selected variables for standardization. Else, center and scale must be of same length as the number of selected variables. Values in center and scale will be matched to selected variables in the provided order, unless a named vector is given. In this case, names are matched against the names of the selected variables.
- For unstandardize():
center and scale correspond to the center (the mean / median) and the scale (SD / MAD) of the original non-standardized data (for data frames, should be named, or have column order correspond to the numeric column). However, one can also directly provide the original data through reference, from which the center and the scale will be computed (according to robust and two_sd). Alternatively, if the input contains the attributes center and scale (as does the output of standardize()), it will take it from there if the rest of the arguments are absent.
verbose Toggle warnings and messages on or off.
force
select
Logical, if TRUE, forces recoding of factors and character vectors as well.
Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or c(1, 3, 5)),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or \(-1:-3\) ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select =c("Species", "Test")) will just return "Species".
\begin{tabular}{ll} 
exclude & \begin{tabular}{l} 
See select, however, column names matched by the pattern from exclude will \\
be excluded instead of selected. If NULL (the default), excludes no columns.
\end{tabular} \\
remove_na & \begin{tabular}{l} 
How should missing values (NA) be treated: if "none" (default): each col- \\
umn's standardization is done separately, ignoring NAs. Else, rows with NA in \\
the columns selected with select / exclude ("selected") or in all columns \\
("all") are dropped before standardization, and the resulting data frame does \\
not include these cases. \\
Logical or string. If TRUE, standardized variables get new column names (with \\
the suffix "_z") and are appended (column bind) to x, thus returning both the \\
original and the standardized variables. If FALSE, original variables in x will \\
be overwritten by their standardized versions. If a character value, standardized
\end{tabular} \\
variables are appended with new column names (using the defined suffix) to the \\
original data frame.
\end{tabular}

\section*{Value}

The standardized object (either a standardize data frame or a statistical model fitted on standardized data).

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{Note}

When \(x\) is a vector or a data frame with remove_na = "none"), missing values are preserved, so the return value has the same length / number of rows as the original input.

\section*{See Also}

See center() for grand-mean centering of variables, and makepredictcall.dw_transformer() for use in model formulas.
Other transform utilities: normalize(), ranktransform(), rescale(), reverse()
Other standardize: standardize.default()

\section*{Examples}
```

d <- iris[1:4, ]

# vectors

standardise(d\$Petal.Length)

# Data frames

# overwrite

standardise(d, select = c("Sepal.Length", "Sepal.Width"))

# append

standardise(d, select = c("Sepal.Length", "Sepal.Width"), append = TRUE)

# append, suffix

standardise(d, select = c("Sepal.Length", "Sepal.Width"), append = "_std")

# standardizing with reference center and scale

d <- data.frame(
a = c(-2, -1, 0, 1, 2),
b = c(3, 4, 5, 6, 7)
)

# default standardization, based on mean and sd of each variable

standardize(d) \# means are 0 and 5, sd ~ 1.581139

# standardization, based on mean and sd set to the same values

standardize(d, center = c(0, 5), scale = c(1.581, 1.581))

# standardization, mean and sd for each variable newly defined

standardize(d, center = c(3, 4), scale = c(2, 4))

# standardization, taking same mean and sd for each variable

standardize(d, center = 1, scale = 3)

```
standardize.default Re-fit a model with standardized data

\section*{Description}

Performs a standardization of data (z-scoring) using standardize() and then re-fits the model to the standardized data.

Standardization is done by completely refitting the model on the standardized data. Hence, this approach is equal to standardizing the variables before fitting the model and will return a new model object. This method is particularly recommended for complex models that include interactions or transformations (e.g., polynomial or spline terms). The robust (default to FALSE) argument enables a robust standardization of data, based on the median and the MAD instead of the mean and the SD.
```

Usage
\#\# Default S3 method:
standardize(
x,
robust = FALSE,
two_sd = FALSE,
weights = TRUE,
verbose = TRUE,
include_response = TRUE,
)

```

\section*{Arguments}
\(x \quad\) A statistical model.
robust Logical, if TRUE, centering is done by subtracting the median from the variables and dividing it by the median absolute deviation (MAD). If FALSE, variables are standardized by subtracting the mean and dividing it by the standard deviation (SD).
two_sd If TRUE, the variables are scaled by two times the deviation (SD or MAD depending on robust). This method can be useful to obtain model coefficients of continuous parameters comparable to coefficients related to binary predictors, when applied to the predictors (not the outcome) (Gelman, 2008).
weights If TRUE (default), a weighted-standardization is carried out.
verbose Toggle warnings and messages on or off.
include_response
If TRUE (default), the response value will also be standardized. If FALSE, only the predictors will be standardized.
- Note that for GLMs and models with non-linear link functions, the response value will not be standardized, to make re-fitting the model work.
- If the model contains an stats: :offset(), the offset variable(s) will be standardized only if the response is standardized. If two_sd = TRUE, offsets are standardized by one-sd (similar to the response).
- (For mediate models, the include_response refers to the outcome in the y model; m model's response will always be standardized when possible).
... Arguments passed to or from other methods.

\section*{Value}

A statistical model fitted on standardized data

\section*{Generalized Linear Models}

Standardization for generalized linear models (GLM, GLMM, etc) is done only with respect to the predictors (while the outcome remains as-is, unstandardized) - maintaining the interpretability of the coefficients (e.g., in a binomial model: the exponent of the standardized parameter is the OR of a change of 1 SD in the predictor, etc.)

\section*{Dealing with Factors}
standardize(model) or standardize_parameters(model, method = "refit") do not standardize categorical predictors (i.e. factors) / their dummy-variables, which may be a different behaviour compared to other R packages (such as lm.beta) or other software packages (like SPSS). To mimic such behaviours, either use standardize_parameters(model, method = "basic") to obtain posthoc standardized parameters, or standardize the data with standardize (data, force = TRUE) before fitting the model.

\section*{Transformed Variables}

When the model's formula contains transformations (e.g. y \(\sim \exp (X)\) ) the transformation effectively takes place after standardization (e.g., exp \((\operatorname{scale}(X))\) ). Since some transformations are undefined for none positive values, such as \(\log ()\) and \(\operatorname{sqrt}()\), the relevel variables are shifted (post standardization) by \(Z-\min (Z)+1\) or \(Z-\min (Z)\) (respectively).

\section*{See Also}

Other standardize: standardize()

\section*{Examples}
```

model <- lm(Infant.Mortality ~ Education * Fertility, data = swiss)
coef(standardize(model))

```
```

text_format Convenient text formatting functionalities

```

\section*{Description}

Convenience functions to manipulate and format text.

\section*{Usage}
```

text_format(
text,
sep = ", ",
last = " and ",
width = NULL,
enclose = NULL,
)
format_text(
text,
sep = ", ",
last = " and ",

```
```

        width = NULL,
        enclose = NULL,
    )
text_fullstop(text)
text_lastchar(text, n = 1)
text_concatenate(text, sep = ", ", last = " and ", enclose = NULL)
text_paste(text, text2 = NULL, sep = ", ", enclose = NULL, ...)
text_remove(text, pattern = "", ...)
text_wrap(text, width = NULL, ...)

```

\section*{Arguments}
\begin{tabular}{ll}
\begin{tabular}{l} 
text, text2 \\
sep
\end{tabular} & \begin{tabular}{l} 
A character string. \\
last
\end{tabular} \\
width & \begin{tabular}{l} 
Separator.
\end{tabular} \\
Last separator. \\
enclose & \begin{tabular}{l} 
Positive integer giving the target column width for wrapping lines in the output. \\
Can be "auto", in which case it will select \(90 \backslash\) default width.
\end{tabular} \\
& \begin{tabular}{l} 
Character that will be used to wrap elements of text, so these can be, e.g., \\
enclosed with quotes or backticks. If NULL (default), text elements will not be \\
enclosed.
\end{tabular} \\
\(\ldots\) & \begin{tabular}{l} 
Other arguments to be passed to or from other functions.
\end{tabular} \\
n & \begin{tabular}{l} 
The number of characters to find.
\end{tabular} \\
pattern & \begin{tabular}{l} 
Character vector. For data_rename(), indicates columns that should be se- \\
lected for renaming. Can be NULL (in which case all columns are selected). For \\
data_addprefix() or data_addsuffix(), a character string, which will be \\
added as prefix or suffix to the column names.
\end{tabular}
\end{tabular}

\section*{Value}

A character string.

\section*{Examples}
```


# Add full stop if missing

text_fullstop(c("something", "something else."))

# Find last characters

text_lastchar(c("ABC", "DEF"), n = 2)

# Smart concatenation

text_concatenate(c("First", "Second", "Last"))

```
```

text_concatenate(c("First", "Second", "Last"), last = " or ", enclose = "`")

# Remove parts of string

text_remove(c("one!", "two", "three!"), "!")

# Wrap text

long_text <- paste(rep("abc ", 100), collapse = "")
cat(text_wrap(long_text, width = 50))

# Paste with optional separator

text_paste(c("A", "", "B"), c("42", "42", "42"))

```
```

to_factor Convert data to factors

```

\section*{Description}

Convert data to factors

\section*{Usage}
```

to_factor(x, ...)

## S3 method for class 'numeric'

to_factor(x, labels_to_levels = TRUE, verbose = TRUE, ...)

## S3 method for class 'data.frame'

to_factor(
x,
select = NULL,
exclude = NULL,
ignore_case = FALSE,
append = FALSE,
regex = FALSE,
verbose = TRUE,
...
)

```

\section*{Arguments}
...
labels_to_levels
Logical, if TRUE, value labels are used as factor levels after \(x\) was converted to factor. Else, factor levels are based on the values of \(x\) (i.e. as if using as.factor()).
verbose Toggle warnings.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex(""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is.numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean \((x)>3\) ),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select = c("Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
append Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to \(x\), thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in \(x\) will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
regex
Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length \(>1\). regex \(=\) TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.

\section*{Details}

Convert variables or data into factors. If the data is labelled, value labels will be used as factor levels. The counterpart to convert variables into numeric is to_numeric().

\section*{Value}

A factor, or a data frame of factors.

\section*{Selection of variables - the select argument}

For most functions that have a select argument (including this function), the complete input data frame is returned, even when select only selects a range of variables. That is, the function is only applied to those variables that have a match in select, while all other variables remain unchanged. In other words: for this function, select will not omit any non-included variables, so that the returned data frame will include all variables from the input data frame.

\section*{Note}

Factors are ignored and returned as is. If you want to use value labels as levels for factors, use labels_to_levels() instead.

\section*{Examples}
```

str(to_factor(iris))

# use labels as levels

data(efc)
str(efc$c172code)
head(to_factor(efc$c172code))

```
```

to_numeric Convert data to numeric

```

\section*{Description}

Convert data to numeric by converting characters to factors and factors to either numeric levels or dummy variables. The "counterpart" to convert variables into factors is to_factor().

\section*{Usage}
to_numeric (x, ...)
\#\# S3 method for class 'data.frame'
to_numeric(
x ,
select = NULL,
exclude \(=\) NULL,
dummy_factors = TRUE,
```

    preserve_levels = FALSE,
    lowest = NULL,
    append = FALSE,
    ignore_case = FALSE,
    regex = FALSE,
    verbose = TRUE,
    ...
    )

```

\section*{Arguments}
x
... Arguments passed to or from other methods.
select Variables that will be included when performing the required tasks. Can be either
- a variable specified as a literal variable name (e.g., column_name),
- a string with the variable name (e.g., "column_name"), or a character vector of variable names (e.g., c("col1", "col2", "col3")),
- a formula with variable names (e.g., ~column_1 + column_2),
- a vector of positive integers, giving the positions counting from the left (e.g. 1 or \(c(1,3,5)\) ),
- a vector of negative integers, giving the positions counting from the right (e.g., -1 or -1 : -3 ),
- one of the following select-helpers: starts_with(), ends_with(), contains(), a range using : or regex (""). starts_with(), ends_with(), and contains() accept several patterns, e.g starts_with("Sep", "Petal").
- or a function testing for logical conditions, e.g. is.numeric() (or is .numeric), or any user-defined function that selects the variables for which the function returns TRUE (like: foo <- function(x) mean (x) > 3),
- ranges specified via literal variable names, select-helpers (except regex()) and (user-defined) functions can be negated, i.e. return non-matching elements, when prefixed with a -, e.g. -ends_with(""), -is.numeric or -(Sepal.Width:Petal.Length). Note: Negation means that matches are excluded, and thus, the exclude argument can be used alternatively. For instance, select=-ends_with("Length") (with -) is equivalent to exclude=ends_with("Length") (no -). In case negation should not work as expected, use the exclude argument instead.
If NULL, selects all columns. Patterns that found no matches are silently ignored, e.g. extract_column_names(iris, select \(=c(\) "Species", "Test")) will just return "Species".
exclude See select, however, column names matched by the pattern from exclude will be excluded instead of selected. If NULL (the default), excludes no columns.
dummy_factors Transform factors to dummy factors (all factor levels as different columns filled with a binary \(0-1\) value).
```

preserve_levels
Logical, only applies if $x$ is a factor. If TRUE, and $x$ has numeric factor levels, these will be converted into the related numeric values. If this is not possible, the converted numeric values will start from 1 to number of levels.
lowest Numeric, indicating the lowest (minimum) value when converting factors or character vectors to numeric values.
append Logical or string. If TRUE, recoded or converted variables get new column names and are appended (column bind) to $x$, thus returning both the original and the recoded variables. The new columns get a suffix, based on the calling function: "_r" for recode functions, "_n" for to_numeric(), "_f" for to_factor(), or "_s" for slide(). If append=FALSE, original variables in $x$ will be overwritten by their recoded versions. If a character value, recoded variables are appended with new column names (using the defined suffix) to the original data frame.
ignore_case Logical, if TRUE and when one of the select-helpers or a regular expression is used in select, ignores lower/upper case in the search pattern when matching against variable names.
regex Logical, if TRUE, the search pattern from select will be treated as regular expression. When regex = TRUE, select must be a character string (or a variable containing a character string) and is not allowed to be one of the supported select-helpers or a character vector of length $>1$. regex $=$ TRUE is comparable to using one of the two select-helpers, select = contains("") or select = regex (""), however, since the select-helpers may not work when called from inside other functions (see 'Details'), this argument may be used as workaround.
verbose Toggle warnings.

```

\section*{Value}

A data frame of numeric variables.

\section*{Selection of variables - select argument}

For most functions that have a select argument the complete input data frame is returned, even when select only selects a range of variables. However, for to_numeric(), factors might be converted into dummies, thus, the number of variables of the returned data frame no longer match the input data frame. Hence, when select is used, only those variables (or their dummies) specified in select will be returned. Use append=TRUE to also include the original variables in the returned data frame.

\section*{Note}

By default, to_numeric() converts factors into "binary" dummies, i.e. each factor level is converted into a separate column filled with a binary \(0-1\) value. If only one column is required, use dummy_factors \(=\) FALSE. If you want to preserve the original factor levels (in case these represent numeric values), use preserve_levels = TRUE.

\section*{Examples}
to_numeric(head(ToothGrowth))
```

to_numeric(head(ToothGrowth), dummy_factors = FALSE)

# factors

x <- as.factor(mtcars\$gear)
to_numeric(x, dummy_factors = FALSE)
to_numeric(x, dummy_factors = FALSE, preserve_levels = TRUE)

# same as:

coerce_to_numeric(x)

```

\section*{visualisation_recipe Prepare objects for visualisation}

\section*{Description}

This function prepares objects for visualisation by returning a list of layers with data and geoms that can be easily plotted using for instance ggplot2.

If the see package is installed, the call to visualization_recipe() can be replaced by plot(), which will internally call the former and then plot it using ggplot. The resulting plot can be customized ad-hoc (by adding ggplot's geoms, theme or specifications), or via some of the arguments of visualisation_recipe() that control the aesthetic parameters.

See the specific documentation page for your object's class:
- modelbased: https://easystats.github.io/modelbased/reference/visualisation_recipe. estimate_predicted.html
- correlation: https://easystats.github.io/correlation/reference/visualisation_recipe. easycormatrix.html

\section*{Usage}
visualisation_recipe(x, ...)

\section*{Arguments}
x
An easystats object.
. . Other arguments passed to other functions.
```

weighted_mean Weighted Mean, Median, SD, and MAD

```

\section*{Description}

Weighted Mean, Median, SD, and MAD

\section*{Usage}
weighted_mean(x, weights = NULL, remove_na = TRUE, verbose = TRUE, ...)
weighted_median(x, weights = NULL, remove_na = TRUE, verbose = TRUE, ...)
weighted_sd(x, weights = NULL, remove_na = TRUE, verbose = TRUE, ...)
weighted_mad(
x ,
weights = NULL,
constant \(=1.4826\),
remove_na = TRUE,
verbose = TRUE,
)

\section*{Arguments}

X
weights
remove_na
verbose Show warning when weights are negative?
... arguments to be passed to or from methods.
constant scale factor.

\section*{Examples}
```


## GPA from Siegel }199

x <- c(3.7, 3.3, 3.5, 2.8)
wt <- c(5, 5, 4, 1) / 15
weighted_mean(x, wt)
weighted_median(x, wt)
weighted_sd(x, wt)
weighted_mad(x, wt)

```
```

winsorize Winsorize data

```

\section*{Description}

Winsorize data

\section*{Usage}
```

    winsorize(data, ...)
    ## S3 method for class 'numeric'
    winsorize(
        data,
        threshold = 0.2,
        method = "percentile",
        robust = FALSE,
        verbose = TRUE,
    )
    ```

\section*{Arguments}
data data frame or vector.
... Currently not used.
threshold The amount of winsorization, depends on the value of method:
- For method = "percentile": the amount to winsorize from each tail. The value of threshold must be between 0 and 0.5 and of length 1 .
- For method = "zscore": the number of SD/MAD-deviations from the mean/median (see robust). The value of threshold must be greater than 0 and of length 1.
- For method = "raw": a vector of length 2 with the lower and upper bound for winsorization.
method One of "percentile" (default), "zscore", or "raw".
robust Logical, if TRUE, winsorizing through the "zscore" method is done via the median and the median absolute deviation (MAD); if FALSE, via the mean and the standard deviation.
verbose Not used anymore since datawizard 0.6.6.

\section*{Details}

Winsorizing or winsorization is the transformation of statistics by limiting extreme values in the statistical data to reduce the effect of possibly spurious outliers. The distribution of many statistics can be heavily influenced by outliers. A typical strategy is to set all outliers (values beyond a certain threshold) to a specified percentile of the data; for example, a \(90 \%\) winsorization would see all data
below the 5th percentile set to the 5th percentile, and data above the 95 th percentile set to the 95th percentile. Winsorized estimators are usually more robust to outliers than their more standard forms.

\section*{Value}

A data frame with winsorized columns or a winsorized vector.

\section*{See Also}
- Functions to rename stuff: data_rename(), data_rename_rows(), data_addprefix(), data_addsuffix()
- Functions to reorder or remove columns: data_reorder(), data_relocate(), data_remove()
- Functions to reshape, pivot or rotate data frames: data_to_long(), data_to_wide(), data_rotate()
- Functions to recode data: rescale(), reverse(), categorize(), recode_values(), slide()
- Functions to standardize, normalize, rank-transform: center(), standardize(), normalize(), ranktransform(), winsorize()
- Split and merge data frames: data_partition(), data_merge()
- Functions to find or select columns: data_select(), extract_column_names()
- Functions to filter rows: data_match(), data_filter()

\section*{Examples}
```

hist(iris$Sepal.Length, main = "Original data")
hist(winsorize(iris$Sepal.Length, threshold = 0.2),
xlim = c(4, 8), main = "Percentile Winsorization"
)
hist(winsorize(iris$Sepal.Length, threshold = 1.5, method = "zscore"),
    xlim = c(4, 8), main = "Mean (+/- SD) Winsorization"
)
hist(winsorize(iris$Sepal.Length, threshold = 1.5, method = "zscore", robust = TRUE),
xlim = c(4, 8), main = "Median (+/- MAD) Winsorization"
)
hist(winsorize(iris\$Sepal.Length, threshold = c(5, 7.5), method = "raw"),
xlim = c(4, 8), main = "Raw Thresholds"
)

# Also works on a data frame:

winsorize(iris, threshold = 0.2)

```

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