

Package ‘dataquieR’

March 5, 2025

Title Data Quality in Epidemiological Research

Version 2.5.1

Description Data quality assessments guided by a 'data quality framework introduced by Schmidt and colleagues, 2021' <[doi:10.1186/s12874-021-01252-7](https://doi.org/10.1186/s12874-021-01252-7)> target the data quality dimensions integrity, completeness, consistency, and accuracy. The scope of applicable functions rests on the availability of extensive metadata which can be provided in spreadsheet tables. Either standardized (e.g. as 'html5' reports) or individually tailored reports can be generated. For an introduction into the specification of corresponding metadata, please refer to the 'package website' <https://dataquality.qihs.uni-greifswald.de/VIN_Annotation_of_Metadata.html>.

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URL <https://dataquality.qihs.uni-greifswald.de/>

BugReports <https://gitlab.com/libreumg/dataquieR/-/issues>

Depends R (>= 3.6.0)

Imports dplyr (>= 1.0.2), emmeans, ggplot2 (>= 3.5.0), lme4, lubridate, MASS, MultinomialCI, parallelMap, patchwork (>= 1.3.0), R.devices, rlang, robustbase, qmrparser, utils, rio, readr, scales, withr, lifecycle, units, methods

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acc_cat_distributions *Plots and checks for distributions for categorical variables*

Description

To complete

[Descriptor](#)

Usage

```
acc_cat_distributions(
  resp_vars = NULL,
  group_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable the name of the measurement variable
group_vars	variable the name of the observer, device or reader variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

To complete

Value

A [list](#) with:

- SummaryPlot: [ggplot2::ggplot](#) for the response variable in resp_vars.

See Also

[Online Documentation](#)

acc_distributions *Plots and checks for distributions*

Description

Data quality indicator checks "Unexpected location" and "Unexpected proportion" with histograms.

[Indicator](#)

Usage

```
acc_distributions(  
  resp_vars = NULL,  
  study_data,  
  label_col,  
  item_level = "item_level",  
  check_param = c("any", "location", "proportion"),  
  plot_ranges = TRUE,  
  flip_mode = "noflip",  
  meta_data = item_level,  
  meta_data_v2  
)
```

Arguments

resp_vars	variable list the names of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
check_param	enum any location proportion. Which type of check should be conducted (if possible): a check on the location of the mean or median value of the study data, a check on proportions of categories, or either of them if the necessary metadata is available.
plot_ranges	logical Should the plot show ranges and results from the data quality checks? (default: TRUE)

flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = . . .)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
meta_data	data.frame old name for <code>item_level</code>
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code> .

Value

A [list](#) with:

- `SummaryTable`: [data.frame](#) containing data quality checks for "Unexpected location" (`FLG_acc_ud_loc`) and "Unexpected proportion" (`FLG_acc_ud_prop`) for each response variable in `resp_vars`.
- `SummaryData`: a [data.frame](#) containing data quality checks for "Unexpected location" and / or "Unexpected proportion" for a report
- `SummaryPlotList`: [list](#) of [ggplot2::ggplots](#) for each response variable in `resp_vars`.

Algorithm of this implementation:

- If no response variable is defined, select all variables of type float or integer in the study data.
- Remove missing codes from the study data (if defined in the metadata).
- Remove measurements deviating from (hard) limits defined in the metadata (if defined).
- Exclude variables containing only NA or only one unique value (excluding NAs).
- Perform check for "Unexpected location" if defined in the metadata (needs a `LOCATION_METRIC` (mean or median) and `LOCATION_RANGE` (range of expected values for the mean and median, respectively)).
- Perform check for "Unexpected proportion" if defined in the metadata (needs `PROPOR-TION_RANGE` (range of expected values for the proportions of the categories)).
- Plot histogram(s).

See Also

[Online Documentation](#)

 acc_distributions_ecdf

ECDF plots for distribution checks

Description

Data quality indicator checks "Unexpected location" and "Unexpected proportion" if a grouping variable is included: Plots of empirical cumulative distributions for the subgroups.

Descriptor

Usage

```
acc_distributions_ecdf(
  resp_vars = NULL,
  group_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  n_group_max = getOption("dataquieR.max_group_var_levels_in_plot",
    dataquieR.max_group_var_levels_in_plot_default),
  n_obs_per_group_min = getOption("dataquieR.min_obs_per_group_var_in_plot",
    dataquieR.min_obs_per_group_var_in_plot_default)
)
```

Arguments

resp_vars	variable list the names of the measurement variables
group_vars	variable list the name of the observer, device or reader variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
n_group_max	maximum number of categories to be displayed individually for the grouping variable (group_vars, devices / examiners)
n_obs_per_group_min	minimum number of data points per group to create a graph for an individual category of the group_vars variable

Value

A [list](#) with:

- SummaryPlotList: [list](#) of [ggplot2::ggplots](#) for each response variable in resp_vars.

See Also

[Online Documentation](#)

acc_distributions_loc *Plots and checks for distributions – Location*

Description

Data quality indicator checks "Unexpected location" and "Unexpected proportion" with histograms.

[Indicator](#)

Usage

```
acc_distributions_loc(
  resp_vars = NULL,
  study_data,
  label_col = VAR_NAMES,
  item_level = "item_level",
  check_param = "location",
  plot_ranges = TRUE,
  flip_mode = "noflip",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable list the names of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
check_param	enum any location proportion. Which type of check should be conducted (if possible): a check on the location of the mean or median value of the study data, a check on proportions of categories, or either of them if the necessary metadata is available.
plot_ranges	logical Should the plot show ranges and results from the data quality checks? (default: TRUE)

flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = . . .)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
meta_data	data.frame old name for <code>item_level</code>
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code> .

Value

A [list](#) with:

- **SummaryTable**: [data.frame](#) containing data quality checks for "Unexpected location" (`FLG_acc_ud_loc`) and "Unexpected proportion" (`FLG_acc_ud_prop`) for each response variable in `resp_vars`.
- **SummaryData**: a [data.frame](#) containing data quality checks for "Unexpected location" and / or "Unexpected proportion" for a report
- **SummaryPlotList**: [list](#) of [ggplot2::ggplots](#) for each response variable in `resp_vars`.

Algorithm of this implementation:

- If no response variable is defined, select all variables of type float or integer in the study data.
- Remove missing codes from the study data (if defined in the metadata).
- Remove measurements deviating from (hard) limits defined in the metadata (if defined).
- Exclude variables containing only NA or only one unique value (excluding NAs).
- Perform check for "Unexpected location" if defined in the metadata (needs a `LOCATION_METRIC` (mean or median) and `LOCATION_RANGE` (range of expected values for the mean and median, respectively)).
- Perform check for "Unexpected proportion" if defined in the metadata (needs `PROPORTION_RANGE` (range of expected values for the proportions of the categories)).
- Plot histogram(s).

See Also

- [acc_distributions](#)
- [Online Documentation](#)

 acc_distributions_only

Plots and checks for distributions – only

Description

Descriptor

Usage

```
acc_distributions_only(
  resp_vars = NULL,
  study_data,
  label_col = VAR_NAMES,
  item_level = "item_level",
  flip_mode = "noflip",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable list the names of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = ...)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
meta_data	data.frame old name for <code>item_level</code>
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code> .

Value

A [list](#) with:

- `SummaryTable`: [data.frame](#) containing data quality checks for "Unexpected location" (FLG_acc_ud_loc) and "Unexpected proportion" (FLG_acc_ud_prop) for each response variable in `resp_vars`.
- `SummaryData`: a [data.frame](#) containing data quality checks for "Unexpected location" and / or "Unexpected proportion" for a report
- `SummaryPlotList`: [list](#) of `ggplot2::ggplots` for each response variable in `resp_vars`.

Algorithm of this implementation:

- If no response variable is defined, select all variables of type float or integer in the study data.
- Remove missing codes from the study data (if defined in the metadata).
- Remove measurements deviating from (hard) limits defined in the metadata (if defined).
- Exclude variables containing only NA or only one unique value (excluding NAs).
- Perform check for "Unexpected location" if defined in the metadata (needs a LOCATION_METRIC (mean or median) and LOCATION_RANGE (range of expected values for the mean and median, respectively)).
- Perform check for "Unexpected proportion" if defined in the metadata (needs PROPORTION_RANGE (range of expected values for the proportions of the categories)).
- Plot histogram(s).

See Also

- [acc_distributions](#)
- [Online Documentation](#)

acc_distributions_prop

Plots and checks for distributions – Proportion

Description

Data quality indicator checks "Unexpected location" and "Unexpected proportion" with histograms.

[Indicator](#)

Usage

```
acc_distributions_prop(  
  resp_vars = NULL,  
  study_data,  
  label_col,  
  item_level = "item_level",  
  check_param = "proportion",  
  plot_ranges = TRUE,  
  flip_mode = "noflip",  
  meta_data = item_level,  
  meta_data_v2  
)
```

Arguments

resp_vars	variable list the names of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
check_param	enum any location proportion. Which type of check should be conducted (if possible): a check on the location of the mean or median value of the study data, a check on proportions of categories, or either of them if the necessary metadata is available.
plot_ranges	logical Should the plot show ranges and results from the data quality checks? (default: TRUE)
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = ...)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
meta_data	data.frame old name for <code>item_level</code>
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code> .

Value

A [list](#) with:

- `SummaryTable`: [data.frame](#) containing data quality checks for "Unexpected location" (FLG_acc_ud_loc) and "Unexpected proportion" (FLG_acc_ud_prop) for each response variable in `resp_vars`.
- `SummaryData`: a [data.frame](#) containing data quality checks for "Unexpected location" and / or "Unexpected proportion" for a report
- `SummaryPlotList`: [list](#) of [ggplot2::ggplots](#) for each response variable in `resp_vars`.

Algorithm of this implementation:

- If no response variable is defined, select all variables of type float or integer in the study data.
- Remove missing codes from the study data (if defined in the metadata).
- Remove measurements deviating from (hard) limits defined in the metadata (if defined).
- Exclude variables containing only NA or only one unique value (excluding NAs).
- Perform check for "Unexpected location" if defined in the metadata (needs a `LOCATION_METRIC` (mean or median) and `LOCATION_RANGE` (range of expected values for the mean and median, respectively)).
- Perform check for "Unexpected proportion" if defined in the metadata (needs `PROPORTION_RANGE` (range of expected values for the proportions of the categories)).
- Plot histogram(s).

See Also

- [acc_distributions](#)
- [Online Documentation](#)

acc_end_digits	<i>Extension of acc_shape_or_scale to examine uniform distributions of end digits</i>
----------------	---

Description

This implementation contrasts the empirical distribution of a measurement variables against assumed distributions. The approach is adapted from the idea of rootograms (Tukey (1977)) which is also applicable for count data (Kleiber and Zeileis (2016)).

Indicator**Usage**

```
acc_end_digits(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable the names of the measurement variables, mandatory
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Value

a [list](#) with:

- SummaryTable: [data.frame](#) with the columns Variables and FLG_acc_ud_shape
- SummaryPlot: ggplot2 distribution plot comparing expected with observed distribution

ALGORITHM OF THIS IMPLEMENTATION:

- This implementation is restricted to data of type float or integer.
- Missing codes are removed from resp_vars (if defined in the metadata)
- The user must specify the column of the metadata containing probability distribution (currently only: normal, uniform, gamma)
- Parameters of each distribution can be estimated from the data or are specified by the user
- A histogram-like plot contrasts the empirical vs. the technical distribution

See Also

[Online Documentation](#)

acc_loess	<i>Smooths and plots adjusted longitudinal measurements and longitudinal trends from logistic regression models</i>
-----------	---

Description

The following R implementation executes calculations for quality indicator "Unexpected location" (see [here](#)). Local regression (LOESS) is a versatile statistical method to explore an averaged course of time series measurements (Cleveland, Devlin, and Grosse 1988). In context of epidemiological data, repeated measurements using the same measurement device or by the same examiner can be considered a time series. LOESS allows to explore changes in these measurements over time.

[Descriptor](#)

Usage

```
acc_loess(
  resp_vars,
  group_vars = NULL,
  time_vars,
  co_vars = NULL,
  study_data,
  label_col = VAR_NAMES,
  item_level = "item_level",
  min_obs_in_subgroup = 30,
  resolution = 80,
  comparison_lines = list(type = c("mean/sd", "quartiles"), color = "grey30", linetype =
    2, sd_factor = 0.5),
  mark_time_points = getOption("dataquieR.acc_loess.mark_time_points",
    dataquieR.acc_loess.mark_time_points_default),
  plot_observations = getOption("dataquieR.acc_loess.plot_observations",
    dataquieR.acc_loess.plot_observations_default),
  plot_format = getOption("dataquieR.acc_loess.plot_format",
    dataquieR.acc_loess.plot_format_default),
```

```

meta_data = item_level,
meta_data_v2,
n_group_max = getOption("dataquieR.max_group_var_levels_in_plot",
  dataquieR.max_group_var_levels_in_plot_default),
enable_GAM = getOption("dataquieR.GAM_for_LOESS", dataquieR.GAM_for_LOESS.default),
exclude_constant_subgroups =
  getOption("dataquieR.acc_loess.exclude_constant_subgroups",
    dataquieR.acc_loess.exclude_constant_subgroups.default),
min_bandwidth = getOption("dataquieR.acc_loess.min_bw",
  dataquieR.acc_loess.min_bw.default),
min_proportion = getOption("dataquieR.acc_loess.min_proportion",
  dataquieR.acc_loess.min_proportion.default)
)

```

Arguments

resp_vars	variable the name of the continuous measurement variable
group_vars	variable the name of the observer, device or reader variable
time_vars	variable the name of the variable giving the time of measurement
co_vars	variable list a vector of covariables for adjustment, for example age and sex. Can be NULL (default) for no adjustment.
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
min_obs_in_subgroup	integer (optional argument) If group_vars is specified, this argument can be used to specify the minimum number of observations required for each of the subgroups. Subgroups with fewer observations are excluded. The default number is 30.
resolution	numeric the maximum number of time points used for plotting the trend lines
comparison_lines	list type and style of lines with which trend lines are to be compared. Can be mean +/- 0.5 standard deviation (the factor can be specified differently in sd_factor) or quartiles (Q1, Q2, and Q3). Arguments color and linetype are passed to <code>ggplot2::geom_line()</code> .
mark_time_points	logical mark time points with observations (caution, there may be many marks)
plot_observations	logical show observations as scatter plot in the background. If there are co_vars specified, the values of the observations in the plot will also be adjusted for the specified covariables.
plot_format	enum AUTO COMBINED FACETS BOTH. Return the plot as one combined plot for all groups or as facet plots (one figure per group). BOTH will return both variants, AUTO will decide based on the number of observers.
meta_data	data.frame old name for item_level

meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
n_group_max	integer maximum number of categories to be displayed individually for the grouping variable (group_vars, devices / examiners)
enable_GAM	logical Can LOESS computations be replaced by general additive models to reduce memory consumption for large datasets?
exclude_constant_subgroups	logical Should subgroups with constant values be excluded?
min_bandwidth	numeric lower limit for the LOESS bandwidth, should be greater than 0 and less than or equal to 1. In general, increasing the bandwidth leads to a smoother trend line.
min_proportion	numeric lower limit for the proportion of the smaller group (cases or controls) for creating a LOESS figure, should be greater than 0 and less than 0.4.

Details

If `mark_time_points` or `plot_observations` is selected, but would result in plotting more than 400 points, only a sample of the data will be displayed.

Limitations

The application of LOESS requires model fitting, i.e. the smoothness of a model is subject to a smoothing parameter (span). Particularly in the presence of interval-based missing data, high variability of measurements combined with a low number of observations in one level of the `group_vars` may distort the fit. Since our approach handles data without knowledge of such underlying characteristics, finding the best fit is complicated if computational costs should be minimal. The default of LOESS in R uses a span of 0.75, which provides in most cases reasonable fits. The function `acc_loess` adapts the span for each level of the `group_vars` (with at least as many observations as specified in `min_obs_in_subgroup` and with at least three time points) based on the respective number of observations. LOESS consumes a lot of memory for larger datasets. That is why `acc_loess` switches to a generalized additive model with integrated smoothness estimation (gam by `mgcv`) if there are 1000 observations or more for at least one level of the `group_vars` (similar to `geom_smooth` from `ggplot2`).

Value

a **list** with:

- `SummaryPlotList`: list with two plots if `plot_format = "BOTH"`, otherwise one of the two figures described below:
 - `Loess_fits_facets`: The plot contains LOESS-smoothed curves for each level of the `group_vars` in a separate panel. Added trend lines represent mean and standard deviation or quartiles (specified in `comparison_lines`) for moving windows over the whole data.
 - `Loess_fits_combined`: This plot combines all curves into one panel. Given a low number of levels in the `group_vars`, this plot eases comparisons. However, if the number increases this plot may be too crowded and unclear.

See Also

[Online Documentation](#)

acc_margins

Estimate marginal means, see `emmeans::emmeans`

Description

This function examines the impact of so-called process variables on a measurement variable. This implementation combines a descriptive and a model-based approach. Process variables that can be considered in this implementation must be categorical. It is currently not possible to consider more than one process variable within one function call. The measurement variable can be adjusted for (multiple) covariables, such as age or sex, for example.

Marginal means rests on model-based results, i.e. a significantly different marginal mean depends on sample size. Particularly in large studies, small and irrelevant differences may become significant. The contrary holds if sample size is low.

Indicator**Usage**

```
acc_margins(  
  resp_vars = NULL,  
  group_vars = NULL,  
  co_vars = NULL,  
  study_data,  
  label_col,  
  item_level = "item_level",  
  threshold_type = "empirical",  
  threshold_value,  
  min_obs_in_subgroup = 5,  
  min_obs_in_cat = 5,  
  dichotomize_categorical_resp = TRUE,  
  cut_off_linear_model_for_ord = 10,  
  meta_data = item_level,  
  meta_data_v2,  
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",  
    dataquieR.acc_margins_sort_default),  
  include_numbers_in_figures = getOption("dataquieR.acc_margins_num",  
    dataquieR.acc_margins_num_default),  
  n_violin_max = getOption("dataquieR.max_group_var_levels_with_violins",  
    dataquieR.max_group_var_levels_with_violins_default)  
)
```

Arguments

resp_vars	variable the name of the measurement variable
group_vars	variable list len=1-1. the name of the observer, device or reader variable
co_vars	variable list a vector of covariables, e.g. age and sex for adjustment
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
threshold_type	enum empirical user none. In case empirical is chosen, a multiplier of the scale measure is used. In case of user, a value of the mean or probability (binary data) has to be defined see Implementation and use of thresholds in the online documentation). In case of none, no thresholds are displayed and no flagging of unusual group levels is applied.
threshold_value	numeric a multiplier or absolute value (see Implementation and use of thresholds in the online documentation).
min_obs_in_subgroup	integer from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis. Subgroups with less observations are excluded.
min_obs_in_cat	integer This optional argument specifies the minimum number of observations that is required to include a category (level) of the outcome (resp_vars) in the analysis. Categories with less observations are combined into one group. If the collapsed category contains less observations than required, it will be excluded from the analysis.
dichotomize_categorical_resp	logical Should nominal response variables always be transformed to binary variables?
cut_off_linear_model_for_ord	integer from=0. This optional argument specifies the minimum number of observations for individual levels of an ordinal outcome (resp_var) that is required to run a linear model instead of an ordered regression (i.e., a cut-off value above which linear models are considered a good approximation). The argument can be set to NULL if ordered regression models are preferred for ordinal data in any case.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
sort_group_var_levels	logical Should the levels of the grouping variable be sorted descending by the number of observations? Note that ordinal grouping variables will not be re-ordered.
include_numbers_in_figures	logical Should the figure report the number of observations for each level of the grouping variable?

n_violin_max [integer](#) from=0. This optional argument specifies the maximum number of levels of the group_var for which violin plots will be shown in the figure.

Details

Limitations

Selecting the appropriate distribution is complex. Dozens of continuous, discrete or mixed distributions are conceivable in the context of epidemiological data. Their exact exploration is beyond the scope of this data quality approach. The present function uses the help function [util_dist_selection](#), the assigned SCALE_LEVEL and the DATA_TYPE to discriminate the following cases:

- continuous data
- binary data
- count data with ≤ 20 distinct values
- count data with > 20 distinct values (treated as continuous)
- nominal data
- ordinal data

Continuous data and count data with more than 20 distinct values are analyzed by linear models. Count data with up to 20 distinct values are modeled by a Poisson regression. For binary data, the implementation uses logistic regression. Nominal response variables will either be transformed to binary variables or analyzed by multinomial logistic regression models. The latter option is only available if the argument `dichotomize_categorical_resp` is set to FALSE and if the package `nnet` is installed. The transformation to a binary variable can be user-specified using the metadata columns `RECODE_CASES` and/or `RECODE_CONTROL`. Otherwise, the most frequent category will be assigned to cases and the remaining categories to control. For ordinal response variables, the argument `cut_off_linear_model_for_ord` controls whether the data is analyzed in the same way as continuous data: If every level of the variable has at least as many observations as specified in the argument, the data will be analyzed by a linear model. Otherwise, the data will be modeled by an ordered regression, if the package `ordinal` is installed.

Value

a list with:

- SummaryTable: [data.frame](#) underlying the plot
- ResultData: [data.frame](#)
- SummaryPlot: `ggplot2::ggplot()` margins plot

See Also

[Online Documentation](#)

 acc_multivariate_outlier

Calculate and plot Mahalanobis distances

Description

A standard tool to detect multivariate outliers is the Mahalanobis distance. This approach is very helpful for the interpretation of the plausibility of a measurement given the value of another. In this approach the Mahalanobis distance is used as a univariate measure itself. We apply the same rules for the identification of outliers as in univariate outliers:

- the classical approach from Tukey: $1.5 * IQR$ from the 1st (Q_{25}) or 3rd (Q_{75}) quartile.
- the 3SD approach, i.e. any measurement of the Mahalanobis distance not in the interval of $\bar{x} \pm 3 * \sigma$ is considered an outlier.
- the approach from Hubert for skewed distributions which is embedded in the R package **robustbase**
- a completely heuristic approach named σ -gap.

For further details, please see the vignette for univariate outlier.

[Indicator](#)

Usage

```
acc_multivariate_outlier(
  variable_group = NULL,
  id_vars = NULL,
  label_col = VAR_NAMES,
  study_data,
  item_level = "item_level",
  n_rules = 4,
  max_non_outliers_plot = 10000,
  criteria = c("tukey", "3sd", "hubert", "sigmagap"),
  meta_data = item_level,
  meta_data_v2,
  scale = getOption("dataquieR.acc_multivariate_outlier.scale",
    dataquieR.acc_multivariate_outlier.scale_default),
  multivariate_outlier_check = TRUE
)
```

Arguments

`variable_group` [variable list](#) the names of the continuous measurement variables building a group, for that multivariate outliers make sense.

`id_vars` [variable](#) optional, an ID variable of the study data. If not specified row numbers are used.

`label_col` [variable attribute](#) the name of the column in the metadata with labels of variables

study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
n_rules	numeric from=1 to=4. the no. of rules that must be violated to classify as outlier
max_non_outliers_plot	integer from=0. Maximum number of non-outlier points to be plot. If more points exist, a subsample will be plotted only. Note, that sampling is not deterministic.
criteria	set tukey 3SD hubert sigmagap. a vector with methods to be used for detecting outliers.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
scale	logical Should min-max-scaling be applied per variable?
multivariate_outlier_check	logical really check, pipeline use, only.

Value

a list with:

- SummaryTable: [data.frame](#) underlying the plot
- SummaryPlot: [ggplot2::ggplot2](#) outlier plot
- FlaggedStudyData [data.frame](#) contains the original data frame with the additional columns tukey, 3SD, hubert, and sigmagap. Every observation is coded 0 if no outlier was detected in the respective column and 1 if an outlier was detected. This can be used to exclude observations with outliers.

ALGORITHM OF THIS IMPLEMENTATION:

- Implementation is restricted to variables of type float
- Remove missing codes from the study data (if defined in the metadata)
- The covariance matrix is estimated for all variables from variable_group
- The Mahalanobis distance of each observation is calculated $MD_i^2 = (x_i - \mu)^T \Sigma^{-1} (x_i - \mu)$
- The four rules mentioned above are applied on this distance for each observation in the study data
- An output data frame is generated that flags each outlier
- A parallel coordinate plot indicates respective outliers

List function.

See Also

[Online Documentation](#)

`acc_robust_univariate_outlier`*Identify univariate outliers by four different approaches*

Description

A classical but still popular approach to detect univariate outlier is the boxplot method introduced by Tukey 1977. The boxplot is a simple graphical tool to display information about continuous univariate data (e.g., median, lower and upper quartile). Outliers are defined as values deviating more than $1.5 \times IQR$ from the 1st (Q25) or 3rd (Q75) quartile. The strength of Tukey's method is that it makes no distributional assumptions and thus is also applicable to skewed or non mound-shaped data Marsh and Seo, 2006. Nevertheless, this method tends to identify frequent measurements which are falsely interpreted as true outliers.

A somewhat more conservative approach in terms of symmetric and/or normal distributions is the 3SD approach, i.e. any measurement not in the interval of $mean(x) + / - 3 * \sigma$ is considered an outlier.

Both methods mentioned above are not ideally suited to skewed distributions. As many biomarkers such as laboratory measurements represent in skewed distributions the methods above may be insufficient. The approach of Hubert and Vandervieren 2008 adjusts the boxplot for the skewness of the distribution. This approach is implemented in several R packages such as `robustbase::mc` which is used in this implementation of `dataquieR`.

Another completely heuristic approach is also included to identify outliers. The approach is based on the assumption that the distances between measurements of the same underlying distribution should homogeneous. For comprehension of this approach:

- consider an ordered sequence of all measurements.
- between these measurements all distances are calculated.
- the occurrence of larger distances between two neighboring measurements may than indicate a distortion of the data. For the heuristic definition of a large distance $1 * \sigma$ has been chosen.

Note, that the plots are not deterministic, because they use `ggplot2::geom_jitter`.

Indicator

Usage

```
acc_robust_univariate_outlier(  
  resp_vars = NULL,  
  study_data,  
  label_col,  
  item_level = "item_level",  
  exclude_roles,  
  n_rules = length(unique(criteria)),  
  max_non_outliers_plot = 10000,  
  criteria = c("tukey", "3sd", "hubert", "sigmagap"),
```

```

    meta_data = item_level,
    meta_data_v2
  )

```

Arguments

resp_vars [variable list](#) the name of the continuous measurement variable
study_data [data.frame](#) the data frame that contains the measurements
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
item_level [data.frame](#) the data frame that contains metadata attributes of study data
exclude_roles [variable roles](#) a character (vector) of variable roles not included
n_rules [integer](#) from=1 to=4. the no. rules that must be violated to flag a variable as containing outliers. The default is 4, i.e. all.
max_non_outliers_plot [integer](#) from=0. Maximum number of non-outlier points to be plot. If more points exist, a subsample will be plotted only. Note, that sampling is not deterministic.
criteria [set](#) tukey | 3SD | hubert | sigmagap. a vector with methods to be used for detecting outliers.
meta_data [data.frame](#) old name for item_level
meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Details

Hint: *The function is designed for unimodal data only.*

Value

a list with:

- SummaryTable: [data.frame](#) with the columns Variables, Mean, SD, Median, Skewness, Tukey (N), 3SD (N), Hubert (N), Sigma-gap (N), NUM_acc_ud_outlu, Outliers, low (N), Outliers, high (N) Grading
 - SummaryData: [data.frame](#) with the columns Variables, Mean, SD, Median, Skewness, Tukey (N), 3SD (N), Hubert (N), Sigma-gap (N), Outliers (N), Outliers, low (N), Outliers, high (N)
 - SummaryPlotList: [ggplot2::ggplot](#) univariate outlier plots

ALGORITHM OF THIS IMPLEMENTATION:

- Select all variables of type float in the study data
- Remove missing codes from the study data (if defined in the metadata)
- Remove measurements deviating from limits defined in the metadata

- Identify outliers according to the approaches of Tukey (Tukey 1977), 3SD (Saleem et al. 2021), Hubert (Hubert and Vandervieren 2008), and SigmaGap (heuristic)
- An output data frame is generated which indicates the no. possible outliers, the direction of deviations (Outliers, low; Outliers, high) for all methods and a summary score which sums up the deviations of the different rules
- A scatter plot is generated for all examined variables, flagging observations according to the no. violated rules (step 5).

See Also

[acc_univariate_outlier](#)

acc_shape_or_scale *Compare observed versus expected distributions*

Description

This implementation contrasts the empirical distribution of a measurement variables against assumed distributions. The approach is adapted from the idea of rootograms (Tukey 1977) which is also applicable for count data (Kleiber and Zeileis 2016).

[Indicator](#)

Usage

```
acc_shape_or_scale(
  resp_vars,
  study_data,
  label_col,
  item_level = "item_level",
  dist_col,
  guess,
  par1,
  par2,
  end_digits,
  flip_mode = "noflip",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable the name of the continuous measurement variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data

dist_col	variable attribute the name of the variable attribute in meta_data that provides the expected distribution of a study variable
guess	logical estimate parameters
par1	numeric first parameter of the distribution if applicable
par2	numeric second parameter of the distribution if applicable
end_digits	logical internal use. check for end digits preferences
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the roptions(dataquieR.flip_mode = . . .). If called from dq_report, you can also pass flip_mode to all function calls or set them specifically using specific_args.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Value

a list with:

- ResultData: [data.frame](#) underlying the plot
- SummaryPlot: [ggplot2::ggplot2](#) probability distribution plot
- SummaryTable: [data.frame](#) with the columns Variables and FLG_acc_ud_shape

ALGORITHM OF THIS IMPLEMENTATION:

- This implementation is restricted to data of type float or integer.
- Missing codes are removed from resp_vars (if defined in the metadata)
- The user must specify the column of the metadata containing probability distribution (currently only: normal, uniform, gamma)
- Parameters of each distribution can be estimated from the data or are specified by the user
- A histogram-like plot contrasts the empirical vs. the technical distribution

See Also

[Online Documentation](#)

`acc_univariate_outlier`*Identify univariate outliers by four different approaches*

Description

A classical but still popular approach to detect univariate outlier is the boxplot method introduced by Tukey 1977. The boxplot is a simple graphical tool to display information about continuous univariate data (e.g., median, lower and upper quartile). Outliers are defined as values deviating more than $1.5 \times IQR$ from the 1st (Q25) or 3rd (Q75) quartile. The strength of Tukey's method is that it makes no distributional assumptions and thus is also applicable to skewed or non mound-shaped data Marsh and Seo, 2006. Nevertheless, this method tends to identify frequent measurements which are falsely interpreted as true outliers.

A somewhat more conservative approach in terms of symmetric and/or normal distributions is the 3SD approach, i.e. any measurement not in the interval of $mean(x) + / - 3 * \sigma$ is considered an outlier.

Both methods mentioned above are not ideally suited to skewed distributions. As many biomarkers such as laboratory measurements represent in skewed distributions the methods above may be insufficient. The approach of Hubert and Vandervieren 2008 adjusts the boxplot for the skewness of the distribution. This approach is implemented in several R packages such as `robustbase::mc` which is used in this implementation of `dataquieR`.

Another completely heuristic approach is also included to identify outliers. The approach is based on the assumption that the distances between measurements of the same underlying distribution should homogeneous. For comprehension of this approach:

- consider an ordered sequence of all measurements.
- between these measurements all distances are calculated.
- the occurrence of larger distances between two neighboring measurements may than indicate a distortion of the data. For the heuristic definition of a large distance $1 * \sigma$ has been chosen.

Note, that the plots are not deterministic, because they use `ggplot2::geom_jitter`.

Indicator

Usage

```
acc_univariate_outlier(  
  resp_vars = NULL,  
  study_data,  
  label_col,  
  item_level = "item_level",  
  exclude_roles,  
  n_rules = length(unique(criteria)),  
  max_non_outliers_plot = 10000,  
  criteria = c("tukey", "3sd", "hubert", "sigmagap"),
```

```

    meta_data = item_level,
    meta_data_v2
  )

```

Arguments

resp_vars [variable list](#) the name of the continuous measurement variable
study_data [data.frame](#) the data frame that contains the measurements
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
item_level [data.frame](#) the data frame that contains metadata attributes of study data
exclude_roles [variable roles](#) a character (vector) of variable roles not included
n_rules [integer](#) from=1 to=4. the no. rules that must be violated to flag a variable as containing outliers. The default is 4, i.e. all.
max_non_outliers_plot [integer](#) from=0. Maximum number of non-outlier points to be plot. If more points exist, a subsample will be plotted only. Note, that sampling is not deterministic.
criteria [set](#) tukey | 3SD | hubert | sigmagap. a vector with methods to be used for detecting outliers.
meta_data [data.frame](#) old name for item_level
meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Details

Hint: *The function is designed for unimodal data only.*

Value

a list with:

- SummaryTable: [data.frame](#) with the columns Variables, Mean, SD, Median, Skewness, Tukey (N), 3SD (N), Hubert (N), Sigma-gap (N), NUM_acc_ud_outlu, Outliers, low (N), Outliers, high (N) Grading
 - SummaryData: [data.frame](#) with the columns Variables, Mean, SD, Median, Skewness, Tukey (N), 3SD (N), Hubert (N), Sigma-gap (N), Outliers (N), Outliers, low (N), Outliers, high (N)
 - SummaryPlotList: [ggplot2::ggplot](#) univariate outlier plots

ALGORITHM OF THIS IMPLEMENTATION:

- Select all variables of type float in the study data
- Remove missing codes from the study data (if defined in the metadata)
- Remove measurements deviating from limits defined in the metadata

- Identify outliers according to the approaches of Tukey (Tukey 1977), 3SD (Saleem et al. 2021), Hubert (Hubert and Vandervieren 2008), and SigmaGap (heuristic)
- An output data frame is generated which indicates the no. possible outliers, the direction of deviations (Outliers, low; Outliers, high) for all methods and a summary score which sums up the deviations of the different rules
- A scatter plot is generated for all examined variables, flagging observations according to the no. violated rules (step 5).

See Also

- [acc_robust_univariate_outlier](#)
- [Online Documentation](#)

acc_varcomp	<i>Utility function to compute model-based ICC depending on the (statistical) data type</i>
-------------	---

Description

This function is still under construction. It is designed to run for any statistical data type as follows:

- Variables with only two distinct values will be modeled by mixed effects logistic regression.
- Nominal variables will be transformed to binary variables. This can be user-specified using the metadata columns RECODE_CASES and/or RECODE_CONTROL. Otherwise, the most frequent category will be assigned to cases and the remaining categories to control. As for other binary variables, the ICC will be computed using a mixed effects logistic regression.
- Ordinal variables will be analyzed by linear mixed effects models, if every level of the variable has at least as many observations as specified in the argument `cut_off_linear_model_for_ord`. Otherwise, the data will be modeled by a mixed effects ordered regression, if the package `ordinal` is available.
- Metric variables with integer values are analyzed by linear mixed effects models.
- For variables with data type `float`, the existing implementation `acc_varcomp` is called, which also uses linear mixed effects models.

Indicator

Usage

```
acc_varcomp(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  min_obs_in_subgroup = 10,
```



```

    min_subgroups = 5,
    cut_off_linear_model_for_ord = 10,
    threshold_value = lifecycle::deprecated(),
    meta_data = item_level,
    meta_data_v2
  )

```

Arguments

resp_vars	variable the name of the measurement variable
group_vars	variable the name of the examiner, device or reader variable
co_vars	variable list a vector of covariables, e.g. age and sex, for adjustment
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
min_obs_in_subgroup	integer from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis. Subgroups with less observations are excluded.
min_subgroups	integer from=0. This optional argument specifies the minimum number of subgroups (level) of the group_var that is required to run the analysis. If there are less subgroups, the analysis is not conducted.
cut_off_linear_model_for_ord	integer from=0. This optional argument specifies the minimum number of observations for individual levels of an ordinal outcome (resp_var) that is required to run a linear mixed effects model instead of a mixed effects ordered regression (i.e., a cut-off value above which linear models are considered a good approximation). The argument can be set to NULL if ordered regression models are preferred for ordinal data in any case.
threshold_value	Deprecated.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

Not yet described

Value

The function returns two data frames, 'SummaryTable' and 'SummaryData', that differ only in the names of the columns.

```
as.data.frame.dataquieR_resultset
```

Convert a full dataquieR report to a data.frame

Description

Deprecated

Usage

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

Arguments

x	Deprecated
...	Deprecated

Value

Deprecated

```
as.list.dataquieR_resultset
```

Convert a full dataquieR report to a list

Description

Deprecated

Usage

```
## S3 method for class 'dataquieR_resultset'  
as.list(x, ...)
```

Arguments

x	Deprecated
...	Deprecated

Value

Deprecated

```
as.list.dataquieR_resultset2
      inefficient way to convert a report to a list. try prep\_set\_backend\(\)
```

Description

inefficient way to convert a report to a list. try [prep_set_backend\(\)](#)

Usage

```
## S3 method for class 'dataquieR_resultset2'
as.list(x, ...)
```

Arguments

x	dataquieR_resultset2
...	no used

Value

[list](#)

ASSOCIATION_DIRECTION *Cross-item level metadata attribute name*

Description

The allowable direction of an association. The input is a string that can be either "positive" or "negative".

Usage

```
ASSOCIATION_DIRECTION
```

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other [meta_data_cross](#): [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

ASSOCIATION_FORM	<i>Cross-item level metadata attribute name</i>
------------------	---

Description

The allowable form of association. The string specifies the form based on a selected list.

Usage

ASSOCIATION_FORM

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other [meta_data_cross](#): [ASSOCIATION_DIRECTION](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

ASSOCIATION_METRIC	<i>Cross-item level metadata attribute name</i>
--------------------	---

Description

The metric underlying the association in [ASSOCIATION_RANGE](#). The input is a string that specifies the analysis algorithm to be used.

Usage

ASSOCIATION_METRIC

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other [meta_data_cross](#): [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

ASSOCIATION_RANGE	<i>Cross-item level metadata attribute name</i>
-------------------	---

Description

Specifies the allowable range of an association. The inclusion of the endpoints follows standard mathematical notation using round brackets for open intervals and square brackets for closed intervals. Values must be separated by a semicolon.

Usage

ASSOCIATION_RANGE

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

CHECK_ID	<i>Cross-item level metadata attribute name</i>
----------	---

Description

Specifies the unique IDs for cross-item level metadata records

Usage

CHECK_ID

Format

An object of class character of length 1.

Details

if missing, `dataquieR` will create such IDs

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

CHECK_LABEL	<i>Cross-item level metadata attribute name</i>
-------------	---

Description

Specifies the unique labels for cross-item level metadata records

Usage

CHECK_LABEL

Format

An object of class character of length 1.

Details

if missing, `dataquieR` will create such labels

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

check_table	<i>Data frame with contradiction rules</i>
-------------	--

Description

Two versions exist, the newer one is used by [con_contradictions_redcap](#) and is described [here](#)., the older one used by [con_contradictions](#) is described [here](#).

See Also

[meta_data_cross](#)

CODE_CLASSES	<i>types of value codes</i>
--------------	-----------------------------

Description

types of value codes

Usage

CODE_CLASSES

Format

An object of class list of length 3.

CODE_LIST_TABLE	<i>Default Name of the Table featuring Code Lists</i>
-----------------	---

Description

Default Name of the Table featuring Code Lists

Metadata sheet name containing VALUE_LABEL_TABLES This metadata sheet can contain both value labels of several VALUE_LABEL_TABLE and also Missing and JUMP tables

Usage

CODE_LIST_TABLE

CODE_LIST_TABLE

Format

An object of class character of length 1.

An object of class character of length 1.

CODE_ORDER	<i>Only existence is checked, order not yet used</i>
------------	--

Description

Only existence is checked, order not yet used

Usage

CODE_ORDER

Format

An object of class character of length 1.

com_item_missingness	<i>Summarize missingness columnwise (in variable)</i>
----------------------	---

Description

Item-Missingness (also referred to as item nonresponse (De Leeuw et al. 2003)) describes the missingness of single values, e.g. blanks or empty data cells in a data set. Item-Missingness occurs for example in case a respondent does not provide information for a certain question, a question is overlooked by accident, a programming failure occurs or a provided answer were missed while entering the data.

Indicator**Usage**

```
com_item_missingness(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  show_causes = TRUE,
  cause_label_df,
  include_sysmiss = TRUE,
  threshold_value,
  suppressWarnings = FALSE,
  assume_consistent_codes = TRUE,
  expand_codes = assume_consistent_codes,
  drop_levels = FALSE,
  expected_observations = c("HIERARCHY", "ALL", "SEGMENT"),
  pretty_print = lifecycle::deprecated(),
  meta_data = item_level,
  meta_data_v2
)
```


Arguments

resp_vars	variable list the name of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
show_causes	logical if TRUE, then the distribution of missing codes is shown
cause_label_df	data.frame missing code table. If missing codes have labels the respective data frame can be specified here or in the metadata as assignments, see cause_label_df
include_sysmiss	logical Optional, if TRUE system missingness (NAs) is evaluated in the summary plot
threshold_value	numeric from=0 to=100. a numerical value ranging from 0-100
suppressWarnings	logical warn about consistency issues with missing and jump lists
assume_consistent_codes	logical if TRUE and no labels are given and the same missing/jump code is used for more than one variable, the labels assigned for this code are treated as being the same for all variables.
expand_codes	logical if TRUE, code labels are copied from other variables, if the code is the same and the label is set somewhere
drop_levels	logical if TRUE, do not display unused missing codes in the figure legend.
expected_observations	enum HIERARCHY ALL SEGMENT. If ALL, all observations are expected to comprise all study segments. If SEGMENT, the PART_VAR is expected to point to a variable with values of 0 and 1, indicating whether the variable was expected to be observed for each data row. If HIERARCHY, this is also checked recursively, so, if a variable points to such a participation variable, and that other variable does has also a PART_VAR entry pointing to a variable, the observation of the initial variable is only expected, if both segment variables are 1.
pretty_print	logical deprecated. If you want to have a human readable output, use SummaryData instead of SummaryTable
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Value

a list with:

- SummaryTable: data frame about item missingness per response variable
- SummaryData: data frame about item missingness per response variable formatted for user
- SummaryPlot: ggplot2 heatmap plot, if show_causes was TRUE
- ReportSummaryTable: data frame underlying SummaryPlot

ALGORITHM OF THIS IMPLEMENTATION:

- Lists of missing codes and, if applicable, jump codes are selected from the metadata
- The no. of system missings (NA) in each variable is calculated
- The no. of used missing codes is calculated for each variable
- The no. of used jump codes is calculated for each variable
- Two result dataframes (1: on the level of observations, 2: a summary for each variable) are generated
- *OPTIONAL*: if show_causes is selected, one summary plot for all resp_vars is provided

See Also

[Online Documentation](#)

com_qualified_item_missingness

Compute Indicators for Qualified Item Missingness

Description

[Indicator](#)

Usage

```
com_qualified_item_missingness(
  resp_vars,
  study_data,
  label_col = NULL,
  item_level = "item_level",
  expected_observations = c("HIERARCHY", "ALL", "SEGMENT"),
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable list the name of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
expected_observations	enum HIERARCHY ALL SEGMENT. Report the number of observations expected using the old PART_VAR concept. See com_item_missingness for an explanation.
meta_data	data.frame old name for item_level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Value

A [list](#) with:

- SummaryTable: [data.frame](#) containing data quality checks for "Non-response rate" (PCT_com_qum_nonresp) and "Refusal rate" (PCT_com_qum_refusal) for each response variable in resp_vars.
- SummaryData: a [data.frame](#) containing data quality checks for "Non-response rate" and "Refusal rate" for a report

Examples

```
## Not run:
prep_load_workbook_like_file("inst/extdata/Metadata_example_v3-6.xlsx")
clean <- prep_get_data_frame("item_level")
clean <- subset(clean, `Metadata name` == "Example" &
  !dataquieR:::util_empty(VAR_NAMES))
clean$`Metadata name` <- NULL
clean[, "MISSING_LIST_TABLE"] <- "missing_matchtable1"
prep_add_data_frames(item_level = clean)
clean <- prep_get_data_frame("missing_matchtable1")
clean <- clean[clean$`Metadata name` == "Example", , FALSE]
clean <-
  clean[suppressWarnings(as.character(as.integer(clean$CODE_VALUE)) ==
    as.character(clean$CODE_VALUE)), , FALSE]
clean$CODE_VALUE <- as.integer(clean$CODE_VALUE)
clean <- clean[!is.na(clean$`Metadata name`), , FALSE]
clean$`Metadata name` <- NULL
prep_add_data_frames(missing_matchtable1 = clean)
ship <- prep_get_data_frame("ship")
number_of_mis <- ceiling(nrow(ship) / 20)
resp_vars <- sample(colnames(ship), ceiling(ncol(ship) / 20), FALSE)
mistab <- prep_get_data_frame("missing_matchtable1")
valid_replacement_codes <-
  mistab[mistab$CODE_INTERPRET != "I", CODE_VALUE,
    drop =
      TRUE] # sample only replacement codes on item level. I uses the actual
    # values
for (rv in resp_vars) {
  values <- sample(as.numeric(valid_replacement_codes), number_of_mis,
    replace = TRUE)
  if (inherits(ship[[rv]], "POSIXct")) {
    values <- as.POSIXct(values, origin = min(as.POSIXct(Sys.Date()), 0))
  }
  ship[sample(seq_len(nrow(ship)), number_of_mis, replace = FALSE), rv] <-
    values
}
com_qualified_item_missingness(resp_vars = NULL, ship, "item_level", LABEL)
com_qualified_item_missingness(resp_vars = "Diabetes Age onset", ship,
```

```

    "item_level", LABEL)
com_qualified_item_missingness(resp_vars = NULL, "study_data", "meta_data",
  LABEL)
study_data <- ship
meta_data <- prep_get_data_frame("item_level")
label <- LABEL

## End(Not run)

```

com_qualified_segment_missingness

Compute Indicators for Qualified Segment Missingness

Description

[Indicator](#)

Usage

```

com_qualified_segment_missingness(
  label_col = NULL,
  study_data,
  item_level = "item_level",
  expected_observations = c("HIERARCHY", "ALL", "SEGMENT"),
  meta_data = item_level,
  meta_data_v2,
  meta_data_segment,
  segment_level
)

```

Arguments

label_col	variable attribute the name of the column in the metadata with labels of variables
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
expected_observations	enum HIERARCHY ALL SEGMENT. Report the number of observations expected using the old PART_VAR concept. See com_item_missingness for an explanation.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
meta_data_segment	data.frame Segment level metadata
segment_level	data.frame alias for meta_data_segment

Value

A [list](#) with:

- SegmentTable: [data.frame](#) containing data quality checks for "Non-response rate" (PCT_com_qum_nonresp) and "Refusal rate" (PCT_com_qum_refusal) for each segment.
- SegmentData: a [data.frame](#) containing data quality checks for "Unexpected location" and "Unexpected proportion" per segment for a report

com_segment_missingness

Summarizes missingness for individuals in specific segments

Description**This implementation can be applied in two use cases::**

1. participation in study segments is not recorded by respective variables, e.g. a participant's refusal to attend a specific examination is not recorded.
2. participation in study segments is recorded by respective variables.

Use case (1) will be common in smaller studies. For the calculation of segment missingness it is assumed that study variables are nested in respective segments. This structure must be specified in the static metadata. The R-function identifies all variables within each segment and returns TRUE if all variables within a segment are missing, otherwise FALSE.

Use case (2) assumes a more complex structure of study data and metadata. The study data comprise so-called intro-variables (either TRUE/FALSE or codes for non-participation). The column PART_VAR in the metadata is filled by variable-IDs indicating for each variable the respective intro-variable. This structure has the benefit that subsequent calculation of item missingness obtains correct denominators for the calculation of missingness rates.

[Descriptor](#)

Usage

```
com_segment_missingness(
  study_data,
  item_level = "item_level",
  strata_vars = NULL,
  group_vars = NULL,
  label_col,
  threshold_value,
  direction,
  color_gradient_direction,
  expected_observations = c("HIERARCHY", "ALL", "SEGMENT"),
  exclude_roles = c(VARIABLE_ROLES$PROCESS),
  meta_data = item_level,
  meta_data_v2,
  segment_level,
  meta_data_segment
)
```

Arguments

study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
strata_vars	variable the name of a variable used for stratification, defaults to NULL for not grouping output
group_vars	variable the name of a variable used for grouping, defaults to <i>NULL</i> for not grouping output
label_col	variable attribute the name of the column in the metadata with labels of variables
threshold_value	numeric from=0 to=100. a numerical value ranging from 0-100
direction	enum low high. "high" or "low", i.e. are deviations above/below the threshold critical. This argument is deprecated and replaced by <i>color_gradient_direction</i> .
color_gradient_direction	enum above below. "above" or "below", i.e. are deviations above or below the threshold critical? (default: above)
expected_observations	enum HIERARCHY ALL SEGMENT. If ALL, all observations are expected to comprise all study segments. If SEGMENT, the PART_VAR is expected to point to a variable with values of 0 and 1, indicating whether the variable was expected to be observed for each data row. If HIERARCHY, this is also checked recursively, so, if a variable points to such a participation variable, and that other variable does has also a PART_VAR entry pointing to a variable, the observation of the initial variable is only expected, if both segment variables are 1.
exclude_roles	variable roles a character (vector) of variable roles not included
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
segment_level	data.frame alias for meta_data_segment
meta_data_segment	data.frame Segment level metadata. Optional.

Details**Implementation and use of thresholds:**

This implementation uses one threshold to discriminate critical from non-critical values. If direction is above than all values below the threshold_value are normal (displayed in dark blue in the plot and flagged with GRADING = 0 in the dataframe). All values above the threshold_value are considered critical. The more they deviate from the threshold the displayed color shifts to dark red. All critical values are highlighted with GRADING = 1 in the summary data frame. By default, highest values are always shown in dark red irrespective of the absolute deviation.

If direction is below than all values above the threshold_value are normal (displayed in dark blue, GRADING = 0).

Hint:

This function does not support a `resp_vars` argument but `exclude_roles` to specify variables not relevant for detecting a missing segment.

List function.

Value

a list with:

- `ResultData`: data frame about segment missingness
- `SummaryPlot`: `ggplot2` heatmap plot: a heatmap-like graphic that highlights critical values depending on the respective `threshold_value` and direction.
- `ReportSummaryTable`: data frame underlying `SummaryPlot`

See Also

[Online Documentation](#)

`com_unit_missingness` *Counts all individuals with no measurements at all*

Description

This implementation examines a crude version of unit missingness or unit-nonresponse (Kalton and Kasprzyk 1986), i.e. if all measurement variables in the study data are missing for an observation it has unit missingness.

The function can be applied on stratified data. In this case `strata_vars` must be specified.

[Descriptor](#)

Usage

```
com_unit_missingness(  
  id_vars = NULL,  
  strata_vars = NULL,  
  label_col,  
  study_data,  
  item_level = "item_level",  
  meta_data = item_level,  
  meta_data_v2  
)
```

Arguments

id_vars	variable list optional, a (vectorized) call of ID-variables that should not be considered in the calculation of unit- missingness
strata_vars	variable optional, a string or integer variable used for stratification
label_col	variable attribute the name of the column in the metadata with labels of variables
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

This implementations calculates a crude rate of unit-missingness. This type of missingness may have several causes and is an important research outcome. For example, unit-nonresponse may be selective regarding the targeted study population or technical reasons such as record-linkage may cause unit-missingness.

It has to be discriminated form segment and item missingness, since different causes and mechanisms may be the reason for unit-missingness.

Hint:

This function does not support a resp_vars argument but id_vars, which have a roughly inverse logic behind: id_vars with values do not prevent a row from being considered missing, because an ID is the only hint for a unit that otherwise would not occur in the data at all.

List function.

Value

A list with:

- FlaggedStudyData: [data.frame](#) with id-only-rows flagged in a column Unit_missing
- SummaryData: [data.frame](#) with numbers and percentages of unit missingness

See Also

[Online Documentation](#)

contradiction_functions_descriptions
description of the contradiction functions

Description

description of the contradiction functions

Usage

contradiction_functions_descriptions

Format

An object of class list of length 11.

CONTRADICTION_TERM *Cross-item level metadata attribute name*

Description

Note: in some prep_-functions, this field is named RULE

Usage

CONTRADICTION_TERM

Format

An object of class character of length 1.

Details

Specifies a contradiction rule. Use REDCap like syntax, see [online vignette](#)

See Also

[meta_data_cross](#)

Other meta_data_cross: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

CONTRADICTION_TYPE *Cross-item level metadata attribute name*

Description

Specifies the type of a contradiction. According to the data quality concept, there are logical and empirical contradictions, see [online vignette](#)

Usage

CONTRADICTION_TYPE

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

con_contradictions *Checks user-defined contradictions in study data*

Description

This approach considers a contradiction if impossible combinations of data are observed in one participant. For example, if age of a participant is recorded repeatedly the value of age is (unfortunately) not able to decline. Most cases of contradictions rest on comparison of two variables.

Important to note, each value that is used for comparison may represent a possible characteristic but the combination of these two values is considered to be impossible. The approach does not consider implausible or inadmissible values.

[Descriptor](#)

Usage

```
con_contradictions(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  threshold_value,
  check_table,
```

```

    summarize_categories = FALSE,
    meta_data = item_level,
    meta_data_v2
  )

```

Arguments

resp_vars [variable list](#) the name of the measurement variables
study_data [data.frame](#) the data frame that contains the measurements
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
item_level [data.frame](#) the data frame that contains metadata attributes of study data
threshold_value [numeric](#) from=0 to=100. a numerical value ranging from 0-100
check_table [data.frame](#) contradiction rules table. Table defining contradictions. See details for its required structure.
summarize_categories [logical](#) Needs a column 'tag' in the check_table. If set, a summary output is generated for the defined categories plus one plot per category.
meta_data [data.frame](#) old name for item_level
meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Details

Algorithm of this implementation::

- Select all variables in the data with defined contradiction rules (static metadata column CONTRADICTIONS)
- Remove missing codes from the study data (if defined in the metadata)
- Remove measurements deviating from limits defined in the metadata
- Assign label to levels of categorical variables (if applicable)
- Apply contradiction checks on predefined sets of variables
- Identification of measurements fulfilling contradiction rules. Therefore two output data frames are generated:
 - on the level of observation to flag each contradictory value combination, and
 - a summary table for each contradiction check.
- A summary plot illustrating the number of contradictions is generated.

List function.

Value

If summarize_categories is FALSE: A [list](#) with:

- FlaggedStudyData: The first output of the contradiction function is a data frame of similar dimension regarding the number of observations in the study data. In addition, for each applied check on the variables an additional column is added which flags observations with a contradiction given the applied check.

- `SummaryTable`: The second output summarizes this information into one data frame. This output can be used to provide an executive overview on the amount of contradictions. This output is meant for automatic digestion within pipelines.
- `SummaryData`: The third output is the same as `SummaryTable` but for human readers.
- `SummaryPlot`: The fourth output visualizes summarized information of `SummaryData`.

if `summarize_categories` is `TRUE`, other objects are returned: one per category named by that category (e.g. "Empirical") containing a result for contradictions within that category only. Additionally, in the slot `all_checks` a result as it would have been returned with `summarize_categories` set to `FALSE`. Finally, a slot `SummaryData` is returned containing sums per Category and an according `ggplot2::ggplot` in `SummaryPlot`.

See Also

[Online Documentation](#)

con_contradictions_redcap

Checks user-defined contradictions in study data

Description

This approach considers a contradiction if impossible combinations of data are observed in one participant. For example, if age of a participant is recorded repeatedly the value of age is (unfortunately) not able to decline. Most cases of contradictions rest on comparison of two variables.

Important to note, each value that is used for comparison may represent a possible characteristic but the combination of these two values is considered to be impossible. The approach does not consider implausible or inadmissible values.

[Indicator](#)

Usage

```
con_contradictions_redcap(
  study_data,
  item_level = "item_level",
  label_col,
  threshold_value,
  meta_data_cross_item = "cross-item_level",
  use_value_labels,
  summarize_categories = FALSE,
  meta_data = item_level,
  cross_item_level,
  `cross-item_level`,
  meta_data_v2
)
```

Arguments

study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
threshold_value	numeric from=0 to=100. a numerical value ranging from 0-100
meta_data_cross_item	data.frame contradiction rules table. Table defining contradictions. See online documentation for its required structure.
use_value_labels	logical Deprecated in favor of DATA_PREPARATION . If set to TRUE, labels can be used in the REDCap syntax to specify contraction checks for categorical variables. If set to FALSE, contractions have to be specified using the coded values. In case that this argument is not set in the function call, it will be set to TRUE if the metadata contains a column VALUE_LABELS which is not empty.
summarize_categories	logical Needs a column CONTRADICTION_TYPE in the meta_data_cross_item. If set, a summary output is generated for the defined categories plus one plot per category. TODO: Not yet controllable by metadata.
meta_data	data.frame old name for item_level
cross_item_level	data.frame alias for meta_data_cross_item
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
'cross-item_level'	data.frame alias for meta_data_cross_item

Details

Algorithm of this implementation::

- Remove missing codes from the study data (if defined in the metadata)
- Remove measurements deviating from limits defined in the metadata
- Assign label to levels of categorical variables (if applicable)
- Apply contradiction checks (given as REDCap-like rules in a separate metadata table)
- Identification of measurements fulfilling contradiction rules. Therefore two output data frames are generated:
 - on the level of observation to flag each contradictory value combination, and
 - a summary table for each contradiction check.
- A summary plot illustrating the number of contradictions is generated.

List function.

Value

If `summarize_categories` is FALSE: A [list](#) with:

- `FlaggedStudyData`: The first output of the contradiction function is a data frame of similar dimension regarding the number of observations in the study data. In addition, for each applied check on the variables an additional column is added which flags observations with a contradiction given the applied check.
- `VariableGroupData`: The second output summarizes this information into one data frame. This output can be used to provide an executive overview on the amount of contradictions.
- `VariableGroupTable`: A subset of `VariableGroupData` used within the pipeline.
- `SummaryPlot`: The third output visualizes summarized information of `SummaryData`.

If `summarize_categories` is TRUE, other objects are returned: A list with one element `Other`, a list with the following entries: One per category named by that category (e.g. "Empirical") containing a result for contradiction checks within that category only. Additionally, in the slot `all_checks`, a result as it would have been returned with `summarize_categories` set to FALSE. Finally, in the top-level list, a slot `SummaryData` is returned containing sums per Category and an according `ggplot2::ggplot` in `SummaryPlot`.

See Also

[Online Documentation for the function `meta_data_cross`](#) [Online Documentation for the required cross-item-level metadata](#)

`con_inadmissible_categorical`

Detects variable levels not specified in metadata

Description

For each categorical variable, value lists should be defined in the metadata. This implementation will examine, if all observed levels in the study data are valid.

[Indicator](#)

Usage

```
con_inadmissible_categorical(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  threshold_value = 0,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable list the name of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
threshold_value	numeric from=0 to=100. a numerical value ranging from 0-100.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details**Algorithm of this implementation::**

- Remove missing codes from the study data (if defined in the metadata)
- Interpretation of variable specific VALUE_LABELS as supplied in the metadata.
- Identification of measurements not corresponding to the expected categories. Therefore two output data frames are generated:
 - on the level of observation to flag each undefined category, and
 - a summary table for each variable.
- Values not corresponding to defined categories are removed in a data frame of modified study data

Value

a list with:

- SummaryData: data frame summarizing inadmissible categories with the columns:
 - Variables: variable name/label
 - OBSERVED_CATEGORIES: the categories observed in the study data
 - DEFINED_CATEGORIES: the categories defined in the metadata
 - NON_MATCHING: the categories observed but not defined
 - NON_MATCHING_N: the number of observations with categories not defined
 - NON_MATCHING_N_PER_CATEGORY: the number of observations for each of the unexpected categories
- SummaryTable: data frame for the dataquieR pipeline reporting the number and percentage of inadmissible categorical values
- ModifiedStudyData: study data having inadmissible categories removed
- FlaggedStudyData: study data having cases with inadmissible categories flagged

See Also

[Online Documentation](#)

con_inadmissible_vocabulary

Detects variable levels not specified in standardized vocabulary

Description

For each categorical variable, value lists should be defined in the metadata. This implementation will examine, if all observed levels in the study data are valid.

Indicator

Usage

```
con_inadmissible_vocabulary(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  threshold_value = 0,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable list the name of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
threshold_value	numeric from=0 to=100. a numerical value ranging from 0-100.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

Algorithm of this implementation::

- Remove missing codes from the study data (if defined in the metadata)
- Interpretation of variable specific VALUE_LABELS as supplied in the metadata.
- Identification of measurements not corresponding to the expected categories. Therefore two output data frames are generated:
 - on the level of observation to flag each undefined category, and
 - a summary table for each variable.
- Values not corresponding to defined categories are removed in a data frame of modified study data

Value

a list with:

- **SummaryData**: data frame summarizing inadmissible categories with the columns:
 - **Variables**: variable name/label
 - **OBSERVED_CATEGORIES**: the categories observed in the study data
 - **DEFINED_CATEGORIES**: the categories defined in the metadata
 - **NON_MATCHING**: the categories observed but not defined
 - **NON_MATCHING_N**: the number of observations with categories not defined
 - **NON_MATCHING_N_PER_CATEGORY**: the number of observations for each of the unexpected categories
 - **GRADING**: indicator TRUE/FALSE if inadmissible categorical values were observed (more than indicated by the `threshold_value`)
- **SummaryTable**: data frame for the dataquieR pipeline reporting the number and percentage of inadmissible categorical values
- **ModifiedStudyData**: study data having inadmissible categories removed
- **FlaggedStudyData**: study data having cases with inadmissible categories flagged

See Also

[Online Documentation](#)

Examples

```
## Not run:
sdt <- data.frame(DIAG = c("B050", "B051", "B052", "B999"),
                 MED0 = c("S01XA28", "N07XX18", "ABC", NA), stringsAsFactors = FALSE)
mdt <- tibble::tribble(
  ~ VAR_NAMES, ~ DATA_TYPE, ~ STANDARDIZED_VOCABULARY_TABLE, ~ SCALE_LEVEL, ~ LABEL,
  "DIAG", "string", "<ICD10>", "nominal", "Diagnosis",
  "MED0", "string", "<ATC>", "nominal", "Medication"
)
con_inadmissible_vocabulary(NULL, sdt, mdt, label_col = LABEL)
prep_load_workbook_like_file("meta_data_v2")
il <- prep_get_data_frame("item_level")
il$STANDARDIZED_VOCABULARY_TABLE[[11]] <- "<ICD10GM>"
il$DATA_TYPE[[11]] <- DATA_TYPES$INTEGER
il$SCALE_LEVEL[[11]] <- SCALE_LEVELS$NOMINAL
prep_add_data_frames(item_level = il)
r <- dq_report2("study_data", dimensions = "con")
r <- dq_report2("study_data", dimensions = "con",
               advanced_options = list(dataquieR.non_disclosure = TRUE))
r

## End(Not run)
```

con_limit_deviations *Detects variable values exceeding limits defined in metadata*

Description

Inadmissible numerical values can be of type integer or float. This implementation requires the definition of intervals in the metadata to examine the admissibility of numerical study data.

This helps identify inadmissible measurements according to hard limits (for multiple variables).

Indicator

Usage

```
con_limit_deviations(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  limits = NULL,
  flip_mode = "noflip",
  return_flagged_study_data = FALSE,
  return_limit_categorical = TRUE,
  meta_data = item_level,
  meta_data_v2,
  show_obs = TRUE
)
```

Arguments

resp_vars	variable list the name of the measurement variables
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
limits	enum HARD_LIMITS SOFT_LIMITS DETECTION_LIMITS. what limits from metadata to check for
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = ...)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
return_flagged_study_data	logical return <code>FlaggedStudyData</code> in the result
return_limit_categorical	logical if TRUE return limit deviations also for categorical variables
meta_data	data.frame old name for <code>item_level</code>

meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
show_obs	logical Should (selected) individual observations be marked in the figure for continuous variables?

Details

Algorithm of this implementation::

- Remove missing codes from the study data (if defined in the metadata)
- Interpretation of variable specific intervals as supplied in the metadata.
- Identification of measurements outside defined limits. Therefore two output data frames are generated:
 - on the level of observation to flag each deviation, and
 - a summary table for each variable.
- A list of plots is generated for each variable examined for limit deviations. The histogram-like plots indicate respective limits as well as deviations.
- Values exceeding limits are removed in a data frame of modified study data

Value

a list with:

- FlaggedStudyData [data.frame](#) related to the study data by a 1:1 relationship, i.e. for each observation is checked whether the value is below or above the limits. Optional, see `return_flagged_study_data`.
- SummaryTable [data.frame](#) summarizing limit deviations for each variable.
- SummaryData [data.frame](#) summarizing limit deviations for each variable for a report.
- SummaryPlotList [list](#) of [ggplot2::ggplots](#) The plots for each variable are either a histogram (continuous) or a barplot (discrete).
- ReportSummaryTable: heatmap-like data frame about limit violations

See Also

- [Online Documentation](#)

dataquieR.acc_loess.exclude_constant_subgroups

Exclude subgroups with constant values from LOESS figure

Description

If this option is set to TRUE, time course plots will only show subgroups with more than one distinct value. This might improve the readability of the figure.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.acc_loess.mark_time_points`

Display time-points in LOESS plots

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.acc_loess.min_bw

Lower limit for the LOESS bandwidth

Description

The value should be greater than 0 and less than or equal to 1. In general, increasing the bandwidth leads to a smoother trend line.

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_constant_subgroups](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logical](#), [dataquieR.debug](#), [dataquieR.des_summary](#), [dataquieR.dontwrapresults](#), [dataquieR.fix_column_type_on_read](#), [dataquieR.flip_mode](#), [dataquieR.force_item_specific_missing_codes](#), [dataquieR.force_label_col](#), [dataquieR.grading_formats](#), [dataquieR.grading_rulesets](#), [dataquieR.guess_missing_codes](#), [dataquieR.lang](#), [dataquieR.max_group_var_levels](#), [dataquieR.max_group_var_levels_with_violins](#), [dataquieR.min_obs_per_group_var_in_plot](#), [dataquieR.non_disclosure](#), [dataquieR.progress_fkt](#), [dataquieR.progress_msg_fkt](#), [dataquieR.scale_level_heuristics](#), [dataquieR.scale_level_heuristics_control_metriclevels](#), [dataquieR.testdebug](#)

dataquieR.acc_loess.min_proportion

Lower limit for the proportion of cases or controls to create a smoothed time trend figure

Description

The value should be greater than 0 and less than 0.4. If the proportion of cases or controls is lower than the specified value, the LOESS figure will not be created for the specified binary outcome.

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_constant_subgroups](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#),

dataquieR.col_con_con_empirical, dataquieR.col_con_con_logical, dataquieR.debug, dataquieR.des_summary,
 dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode,
 dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats,
 dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_level,
 dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot,
 dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_metriclevels,
 dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.acc_loess.plot_format

default for Plot-Format in acc_loess()

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE,
 dataquieR.ELEMENT_MISSMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS,
 dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER,
 dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER,
 dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points,
 dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_observations,
 dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale,
 dataquieR.col_con_con_empirical, dataquieR.col_con_con_logical, dataquieR.debug, dataquieR.des_summary,
 dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode,
 dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats,
 dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_level,
 dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot,
 dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_metriclevels,
 dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.acc_loess.plot_observations

Display observations in LOESS plots

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.acc_margins_num`

Include number of observations for each level of the grouping variable in the 'margins' figure

Description

If this option is set to FALSE, the figures created by `acc_margins` will not include the number of observations for each level of the grouping variable. This can be used to obtain clean static plots.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.acc_margins_sort

Sort levels of the grouping variable in the 'margins' figures

Description

If this option is set to TRUE, the levels of the grouping variable in the figure are sorted in descending order according to the number of observations so that levels with more observations are easier to identify. Otherwise, the original order of the levels is retained.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.acc_multivariate_outlier.scale

Apply min-max scaling in parallel coordinates figure to inspect multivariate outliers

Description

boolean, TRUE or FALSE

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`,

dataquieR.col_con_con_empirical, dataquieR.col_con_con_logical, dataquieR.debug, dataquieR.des_summary,
 dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode,
 dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats,
 dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_level,
 dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot,
 dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_metriclevels,
 dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.col_con_con_empirical

Color for empirical contradictions

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE,
 dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS,
 dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER,
 dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER,
 dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points,
 dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format,
 dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort,
 dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_logical, dataquieR.debug,
 dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read,
 dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col,
 dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes,
 dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins,
 dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt,
 dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeimit,
 dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.col_con_con_logical

Color for logical contradictions

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecodeLimit`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.CONDITIONS_LEVEL_TRHESHOLD`

Log Level

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecodeLimit`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.CONDITIONS_WITH_STACKTRACE

Add stack-trace in condition messages (to be deprecated)

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_html_escaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heu`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.debug`

Call `browser()` on errors

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_html_escaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`,

dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.des_summary_hard_lim_remove

Removal of hard limits from data before calculating descriptive statistics.

Description

can be

- TRUE: values outside hard limits will be removed from the data before calculating descriptive statistics
- FALSE: values outside hard limits will not be removed from the original data

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlEscaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_logi, dataquieR.debug, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.dontwrapresults

Disable automatic post-processing of dataquieR function results

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecode`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.ELEMENT_MISMATCH_CHECKTYPE`

Metadata describes more than the current study data

Description

- none: no check will be provided about the match of variables and records available in the study data and described in the metadata
- exact: There must be a 1:1 match between the study data and metadata regarding data frames and segments variables and records
- subset_u: study data are a subset of metadata. All variables from the study data are expected to be present in the metadata, but one or more variables in the metadata are not expected to be present in the study data. In this case a variable present in the study data but not in the metadata would produce an issue.
- subset_m: metadata are a subset of study data. All variables in the metadata are expected to be present in the study data, but one or more variables in the study data are not expected to be present in the metadata.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`,

dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.ERRORS_WITH_CALLER

Set caller for error conditions (to be deprecated)

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_const, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_logical, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.fix_column_type_on_read

Try to avoid fallback to string columns when reading files

Description

If a file does not feature column data types or features data types cell-based, choose that type which matches the majority of the sampled cells of a column for the column's data type.

Details

This may make you miss data type problems but it could fix them, so `prep_get_data_frame()` works better.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecode`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.flip_mode` *Flip-Mode to Use for figures*

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecode`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.force_item_specific_missing_codes

*Converting **MISSING_LIST/JUMP_LIST** to a **MISSING_LIST_TABLE**
create on list per item*

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISSMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_loglik, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.force_label_col

Control, how the label_col argument is used.

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISSMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_loglik

dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.GAM_for_LOESS

Enable to switch to a general additive model instead of LOESS

Description

If this option is set to TRUE, time course plots will use general additive models (GAM) instead of LOESS when the number of observations exceeds a specified threshold. LOESS computations for large datasets have a high memory consumption.

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_const, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_logical, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.grading_formats

Name of the [data.frame](#) featuring a format for grading-values

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecode`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.grading_rulesets`

Name of the data.frame featuring GRADING_RULESET

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecode`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.guess_missing_codes

Control, if dataquieR tries to guess missing-codes from the study data in absence of metadata

Description

TODO

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_constant_subgroups](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logit](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#), [dataquieR.dontwrapresults](#), [dataquieR.fix_column_type_on_read](#), [dataquieR.flip_mode](#), [dataquieR.force_item_specific_missing_codes](#), [dataquieR.force_label_col](#), [dataquieR.grading_formats](#), [dataquieR.grading_rulesets](#), [dataquieR.lang](#), [dataquieR.max_group_var_levels_in_plot](#), [dataquieR.max_group_var_levels_with_violins](#), [dataquieR.min_obs_per_group_var_in_plot](#), [dataquieR.non_disclosure](#), [dataquieR.progress_fkt](#), [dataquieR.progress_msg_fkt](#), [dataquieR.scale_level_heuristics_control_binaryrecode](#), [dataquieR.scale_level_heuristics_control_metriclevels](#), [dataquieR.testdebug](#)

dataquieR.lang

Language-Suffix for metadata Label-Columns

Description

TODO

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_constant_subgroups](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logit](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#), [dataquieR.dontwrapresults](#),

dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.max_group_var_levels_in_plot

Maximum number of levels of the grouping variable shown individually in figures

Description

If there are more examiners or devices than can be shown individually, they will be collapsed into "other".

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlEscaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_loglik, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodeLimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.max_group_var_levels_with_violins

Maximum number of levels of the grouping variable shown with individual histograms ('violins') in 'margins' figures

Description

If there are more examiners or devices, the figure will be reduced to box-plots to save space.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecodeLimit`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.MAX_LABEL_LEN`

Maximum length for variable labels

Description

All variable labels will be shortened to fit this maximum length. Cannot be larger than 200 for technical reasons.

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryrecodeLimit`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.MAX_VALUE_LABEL_LEN

Maximum length for value labels

Description

value labels are restricted to this length

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_const](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logical](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#), [dataquieR.dontwrapresults](#), [dataquieR.fix_column_type_on_read](#), [dataquieR.flip_mode](#), [dataquieR.force_item_specific_missing_codes](#), [dataquieR.force_label_col](#), [dataquieR.grading_formats](#), [dataquieR.grading_rulesets](#), [dataquieR.guess_missing_codes](#), [dataquieR.lang](#), [dataquieR.max_group_var_level](#), [dataquieR.max_group_var_levels_with_violins](#), [dataquieR.min_obs_per_group_var_in_plot](#), [dataquieR.non_disclosure](#), [dataquieR.progress_fkt](#), [dataquieR.progress_msg_fkt](#), [dataquieR.scale_level_heu](#), [dataquieR.scale_level_heuristics_control_metriclevels](#), [dataquieR.testdebug](#)

dataquieR.MESSAGES_WITH_CALLER

Set caller for message conditions (to be deprecated)

Description

TODO

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_const](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logical](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#), [dataquieR.dontwrapresults](#), [dataquieR.fix_column_type_on_read](#), [dataquieR.flip_mode](#), [dataquieR.force_item_specific_missing_codes](#), [dataquieR.force_label_col](#), [dataquieR.grading_formats](#),

dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodelevelimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.min_obs_per_group_var_in_plot

Minimum number of observations per grouping variable that is required to include an individual level of the grouping variable in a figure

Description

Levels of the grouping variable with fewer observations than specified here will be excluded from the figure.

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_logi, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics_control_binaryrecodelevelimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.MULTIVARIATE_OUTLIER_CHECK

Default availability of multivariate outlier checks in reports

Description

can be

- TRUE: for cross-item_level-groups with MULTIVARIATE_OUTLIER_CHECK empty, do a multivariate outlier check
- FALSE: for cross-item_level-groups with MULTIVARIATE_OUTLIER_CHECK empty, don't do a multivariate outlier check
- "auto": for cross-item_level-groups with MULTIVARIATE_OUTLIER_CHECK empty, do multivariate outlier checks, if there is no entry in the column [CONTRADICTION_TERM](#).

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_const`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logical`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_level`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heu`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

`dataquieR.non_disclosure`

Remove all observation-level-real-data from reports

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_logi`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binaryre`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.progress_fkt

function to call on progress increase

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_binary`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

dataquieR.progress_msg_fkt

function to call on progress message update

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISSMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_htmlescaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`,

dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.scale_level_heuristics_control_binaryreco, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR.scale_level_heuristics_control_binaryrecoelimit

*Number of levels to consider a variable ordinal in absence of
SCALE_LEVEL*

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_LEVEL_TRHESHOLD, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LABEL_LEN, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescaped, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.mark_time_points, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_proportion, dataquieR.acc_loess.plot_format, dataquieR.acc_loess.plot_observations, dataquieR.acc_margins_num, dataquieR.acc_margins_sort, dataquieR.acc_multivariate_outlier.scale, dataquieR.col_con_con_empirical, dataquieR.col_con_con_logi, dataquieR.debug, dataquieR.des_summary_hard_lim_remove, dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_levels_in_plot, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_he, dataquieR.testdebug

dataquieR.scale_level_heuristics_control_metriclevels

*Number of levels to consider a variable metric in absence of
SCALE_LEVEL*

Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_html_escaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.testdebug`

<code>dataquieR.testdebug</code>	<i>Disable all interactively used metadata-based function argument provision</i>
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Description

TODO

See Also

Other options: `dataquieR`, `dataquieR.CONDITIONS_LEVEL_TRHESHOLD`, `dataquieR.CONDITIONS_WITH_STACKTRACE`, `dataquieR.ELEMENT_MISMATCH_CHECKTYPE`, `dataquieR.ERRORS_WITH_CALLER`, `dataquieR.GAM_for_LOESS`, `dataquieR.MAX_LABEL_LEN`, `dataquieR.MAX_VALUE_LABEL_LEN`, `dataquieR.MESSAGES_WITH_CALLER`, `dataquieR.MULTIVARIATE_OUTLIER_CHECK`, `dataquieR.VALUE_LABELS_html_escaped`, `dataquieR.WARNINGS_WITH_CALLER`, `dataquieR.acc_loess.exclude_constant_subgroups`, `dataquieR.acc_loess.mark_time_points`, `dataquieR.acc_loess.min_bw`, `dataquieR.acc_loess.min_proportion`, `dataquieR.acc_loess.plot_format`, `dataquieR.acc_loess.plot_observations`, `dataquieR.acc_margins_num`, `dataquieR.acc_margins_sort`, `dataquieR.acc_multivariate_outlier.scale`, `dataquieR.col_con_con_empirical`, `dataquieR.col_con_con_loglik`, `dataquieR.debug`, `dataquieR.des_summary_hard_lim_remove`, `dataquieR.dontwrapresults`, `dataquieR.fix_column_type_on_read`, `dataquieR.flip_mode`, `dataquieR.force_item_specific_missing_codes`, `dataquieR.force_label_col`, `dataquieR.grading_formats`, `dataquieR.grading_rulesets`, `dataquieR.guess_missing_codes`, `dataquieR.lang`, `dataquieR.max_group_var_levels_in_plot`, `dataquieR.max_group_var_levels_with_violins`, `dataquieR.min_obs_per_group_var_in_plot`, `dataquieR.non_disclosure`, `dataquieR.progress_fkt`, `dataquieR.progress_msg_fkt`, `dataquieR.scale_level_heuristics_control_metriclevels`, `dataquieR.scale_level_heuristics_control_metriclevels`

dataquieR.VALUE_LABELS_htmlescaped

*Assume, all VALUE_LABELS are
Rhref<https://www.w3.org/International/questions/qa-escapesHTML>
escaped*

Description

TODO

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.WARNINGS_WITH_CALLER](#), [dataquieR.acc_loess.exclude_con](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logical](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#), [dataquieR.dontwrapresults](#), [dataquieR.fix_column_type_on_read](#), [dataquieR.flip_mode](#), [dataquieR.force_item_specific_missing_codes](#), [dataquieR.force_label_col](#), [dataquieR.grading_formats](#), [dataquieR.grading_rulesets](#), [dataquieR.guess_missing_codes](#), [dataquieR.lang](#), [dataquieR.max_group_var_level](#), [dataquieR.max_group_var_levels_with_violins](#), [dataquieR.min_obs_per_group_var_in_plot](#), [dataquieR.non_disclosure](#), [dataquieR.progress_fkt](#), [dataquieR.progress_msg_fkt](#), [dataquieR.scale_level_heu](#), [dataquieR.scale_level_heuristics_control_metriclevels](#), [dataquieR.testdebug](#)

dataquieR.WARNINGS_WITH_CALLER

Set caller for warning conditions (to be deprecated)

Description

TODO

See Also

Other options: [dataquieR](#), [dataquieR.CONDITIONS_LEVEL_TRHESHOLD](#), [dataquieR.CONDITIONS_WITH_STACKTRACE](#), [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#), [dataquieR.ERRORS_WITH_CALLER](#), [dataquieR.GAM_for_LOESS](#), [dataquieR.MAX_LABEL_LEN](#), [dataquieR.MAX_VALUE_LABEL_LEN](#), [dataquieR.MESSAGES_WITH_CALLER](#), [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#), [dataquieR.VALUE_LABELS_htmlescaped](#), [dataquieR.acc_loess.exclude_con](#), [dataquieR.acc_loess.mark_time_points](#), [dataquieR.acc_loess.min_bw](#), [dataquieR.acc_loess.min_proportion](#), [dataquieR.acc_loess.plot_format](#), [dataquieR.acc_loess.plot_observations](#), [dataquieR.acc_margins_num](#), [dataquieR.acc_margins_sort](#), [dataquieR.acc_multivariate_outlier.scale](#), [dataquieR.col_con_con_empirical](#), [dataquieR.col_con_con_logical](#), [dataquieR.debug](#), [dataquieR.des_summary_hard_lim_remove](#),

dataquieR.dontwrapresults, dataquieR.fix_column_type_on_read, dataquieR.flip_mode, dataquieR.force_item_specific_missing_codes, dataquieR.force_label_col, dataquieR.grading_formats, dataquieR.grading_rulesets, dataquieR.guess_missing_codes, dataquieR.lang, dataquieR.max_group_var_level, dataquieR.max_group_var_levels_with_violins, dataquieR.min_obs_per_group_var_in_plot, dataquieR.non_disclosure, dataquieR.progress_fkt, dataquieR.progress_msg_fkt, dataquieR.scale_level_heuristics, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

dataquieR_resultset *Internal constructor for the internal class [dataquieR_resultset](#).*

Description

creates an object of the class [dataquieR_resultset](#).

Usage

```
dataquieR_resultset(...)
```

Arguments

... properties stored in the object

Details

The class features the following methods:

- [as.data.frame.dataquieR_resultset](#), * [as.list.dataquieR_resultset](#), * [print.dataquieR_resultset](#), * [summary.dataquieR_resultset](#)

Value

an object of the class [dataquieR_resultset](#).

See Also

[dq_report](#)

dataquieR_resultset2-class
Class [dataquieR_resultset2](#).

Description

Class [dataquieR_resultset2](#).

See Also

[dq_report2](#)

dataquieR_resultset_verify

Verify an object of class dataquieR_resultset

Description

Deprecated

Usage

dataquieR_resultset_verify(...)

Arguments

... Deprecated

Value

Deprecated

DATA_PREPARATION

Cross-item level metadata attribute name

Description

For contradiction rules, the required pre-processing steps that can be given. TODO JM: MISSING_LABEL will not work for non-factor variables

Usage

DATA_PREPARATION

Format

An object of class character of length 1.

Details

LABEL LIMITS MISSING_NA MISSING_LABEL MISSING_INTERPRET

See Also

[meta_data_cross](#)

Other meta_data_cross: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

DATA_TYPES

Data Types

Description

Data Types of Study Data:

In the metadata, the following entries are allowed for the [variable attribute DATA_TYPE](#):

Usage

DATA_TYPES

Format

An object of class `list` of length 4.

Details

- `integer` for integer numbers
- `string` for text/string/character data
- `float` for decimal/floating point numbers
- `datetime` for timepoints

Data Types of Function Arguments:

As function arguments, [dataquieR](#) uses additional type specifications:

- `numeric` is a numerical value ([float](#) or [integer](#)), but it is not an allowed `DATA_TYPE` in the metadata. However, some functions may accept `float` or `integer` for specific function arguments. This is, where we use the term `numeric`.
- `enum` allows one element out of a set of allowed options similar to [match.arg](#)
- `set` allows a subset out of a set of allowed options similar to [match.arg](#) with several `.ok = TRUE`.
- `variable` Function arguments of this type expect a character scalar that specifies one variable using the variable identifier given in the metadata attribute `VAR_NAMES` or, if `label_col` is set, given in the metadata attribute given in that argument. Labels can easily be translated using [prep_map_labels](#)
- `variable list` Function arguments of this type expect a character vector that specifies variables using the variable identifiers given in the metadata attribute `VAR_NAMES` or, if `label_col` is set, given in the metadata attribute given in that argument. Labels can easily be translated using [prep_map_labels](#)

See Also

[integer string](#)

DATA_TYPES_OF_R_TYPE *All available data types, mapped from their respective R types*

Description

All available data types, mapped from their respective R types

Usage

```
DATA_TYPES_OF_R_TYPE
```

Format

An object of class list of length 14.

See Also

[prep_dq_data_type_of](#)

des_scatterplot_matrix
Compute Pairwise Correlations

Description

works on variable groups (cross-item_level), which are expected to show a Pearson correlation

Usage

```
des_scatterplot_matrix(  
  label_col,  
  study_data,  
  item_level = "item_level",  
  meta_data_cross_item = "cross-item_level",  
  meta_data = item_level,  
  meta_data_v2,  
  cross_item_level,  
  `cross-item_level`  
)
```


Arguments

label_col	variable attribute the name of the column in the metadata with labels of variables
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data_cross_item	meta_data_cross
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
cross_item_level	data.frame alias for meta_data_cross_item
'cross-item_level'	data.frame alias for meta_data_cross_item

Details

[Descriptor](#) # TODO: This can be an indicator

Value

a list with the slots:

- SummaryPlotList: for each variable group a [ggplot2::ggplot](#) object with pairwise correlation plots
- SummaryData: table with columns VARIABLE_LIST, cors, max_cor, min_cor
- SummaryTable: like SummaryData, but machine readable and with stable column names.

Examples

```
## Not run:
devtools::load_all()
prep_load_workbook_like_file("meta_data_v2")
des_scatterplot_matrix("study_data")

## End(Not run)
```

des_summary

Compute Descriptive Statistics

Description

generates a descriptive overview of the variables in resp_vars.

[Descriptor](#)

Usage

```

des_summary(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  hard_limits_removal = getOption("dataquieR.des_summary_hard_lim_remove",
    dataquieR.des_summary_hard_lim_remove_default),
  ...
)

```

Arguments

resp_vars	variable the name of the measurement variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
hard_limits_removal	logical if TRUE values outside hard limits are removed from the data before calculating descriptive statistics. The default is FALSE
...	arguments to be passed to all called indicator functions if applicable.

Details

TODO

Value

a [list](#) with:

- SummaryTable: [data.frame](#)
- SummaryData: [data.frame](#)

See Also

[Online Documentation](#)

Examples

```
## Not run:
prep_load_workbook_like_file("meta_data_v2")
xx <- des_summary(study_data = "study_data", meta_data =
  prep_get_data_frame("item_level"))
util_html_table(xx$SummaryData)
util_html_table(des_summary(study_data = prep_get_data_frame("study_data"),
  meta_data = prep_get_data_frame("item_level"))$SummaryData)

## End(Not run)
```

```
des_summary_categorical
```

Compute Descriptive Statistics - categorical variables

Description

generates a descriptive overview of the categorical variables (nominal and ordinal) in resp_vars.

[Descriptor](#)

Usage

```
des_summary_categorical(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  hard_limits_removal = getOption("dataquieR.des_summary_hard_lim_remove",
    dataquieR.des_summary_hard_lim_remove_default),
  ...
)
```

Arguments

resp_vars	variable the name of the categorical measurement variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

hard_limits_removal

logical if TRUE values outside hard limits are removed from the data before calculating descriptive statistics. The default is FALSE

... arguments to be passed to all called indicator functions if applicable.

Details

TODO

Value

a **list** with:

- SummaryTable: [data.frame](#)
- SummaryData: [data.frame](#)

See Also

[Online Documentation](#)

Examples

```
## Not run:
prep_load_workbook_like_file("meta_data_v2")
xx <- des_summary_categorical(study_data = "study_data", meta_data =
  prep_get_data_frame("item_level"))
util_html_table(xx$SummaryData)
util_html_table(des_summary_categorical(study_data = prep_get_data_frame("study_data"),
  meta_data = prep_get_data_frame("item_level"))$SummaryData)

## End(Not run)
```

des_summary_continuous

Compute Descriptive Statistics - continuous variables

Description

generates a descriptive overview of continuous variables (ratio and interval) in resp_vars.

[Descriptor](#)

Usage

```
des_summary_continuous(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  hard_limits_removal = getOption("dataquieR.des_summary_hard_lim_remove",
    dataquieR.des_summary_hard_lim_remove_default),
  ...
)
```

Arguments

`resp_vars` [variable](#) the name of the continuous measurement variable

`study_data` [data.frame](#) the data frame that contains the measurements

`label_col` [variable attribute](#) the name of the column in the metadata with labels of variables

`item_level` [data.frame](#) the data frame that contains metadata attributes of study data

`meta_data` [data.frame](#) old name for `item_level`

`meta_data_v2` [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`hard_limits_removal` [logical](#) if TRUE values outside hard limits are removed from the data before calculating descriptive statistics. The default is FALSE

... arguments to be passed to all called indicator functions if applicable.

Details

TODO

Value

a [list](#) with:

- SummaryTable: [data.frame](#)
- SummaryData: [data.frame](#)

See Also

[Online Documentation](#)

Examples

```
## Not run:
prep_load_workbook_like_file("meta_data_v2")
xx <- des_summary_continuous(study_data = "study_data", meta_data =
                             prep_get_data_frame("item_level"))
util_html_table(xx$SummaryData)
util_html_table(des_summary_continuous(study_data = prep_get_data_frame("study_data"),
                                       meta_data = prep_get_data_frame("item_level"))$SummaryData)

## End(Not run)
```

DF_CODE	<i>Data frame level metadata attribute name</i>
---------	---

Description

Name of the data frame

Usage

DF_CODE

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_ELEMENT_COUNT	<i>Data frame level metadata attribute name</i>
------------------	---

Description

Number of expected data elements in a data frame. [numeric](#). Check only conducted if number entered

Usage

DF_ELEMENT_COUNT

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_ID_REF_TABLE	<i>Data frame level metadata attribute name</i>
-----------------	---

Description

The name of the data frame containing the reference IDs to be compared with the IDs in the study data set.

Usage

DF_ID_REF_TABLE

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_ID_VARS	<i>Data frame level metadata attribute name</i>
------------	---

Description

All variables that are to be used as one single ID variable (combined key) in a data frame.

Usage

DF_ID_VARS

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_NAME

Data frame level metadata attribute name

Description

Name of the data frame

Usage

DF_NAME

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_RECORD_CHECK

Data frame level metadata attribute name

Description

The type of check to be conducted when comparing the reference ID table with the IDs delivered in the study data files.

Usage

DF_RECORD_CHECK

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_RECORD_COUNT	<i>Data frame level metadata attribute name</i>
-----------------	---

Description

Number of expected data records in a data frame. [numeric](#). Check only conducted if number entered

Usage

DF_RECORD_COUNT

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_UNIQUE_ID	<i>Data frame level metadata attribute name</i>
--------------	---

Description

Defines expectancies on the uniqueness of the IDs across the rows of a data frame, or the number of times some ID can be repeated.

Usage

DF_UNIQUE_ID

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

DF_UNIQUE_ROWS	<i>Data frame level metadata attribute name</i>
----------------	---

Description

Specifies whether identical data is permitted across rows in a data frame (excluding ID variables)

Usage

```
DF_UNIQUE_ROWS
```

Format

An object of class character of length 1.

See Also

[meta_data_dataframe](#)

dim.dataquieR_resultset2	<i>Get the dimensions of a dq_report2 result</i>
--------------------------	--

Description

Get the dimensions of a dq_report2 result

Usage

```
## S3 method for class 'dataquieR_resultset2'  
dim(x)
```

Arguments

x a dataquieR_resultset2 result

Value

dimensions

dimensions	<i>Names of DQ dimensions</i>
------------	-------------------------------

Description

a vector of data quality dimensions. The supported dimensions are Completeness, Consistency and Accuracy.

Usage

```
dimensions
```

Format

An object of class character of length 3.

Value

Only a definition, not a function, so no return value

See Also

[Data Quality Concept](#)

dimnames.dataquieR_resultset2	<i>Names of a dataquieR report object (v2.0)</i>
-------------------------------	--

Description

Names of a dataquieR report object (v2.0)

Usage

```
## S3 method for class 'dataquieR_resultset2'
dimnames(x)
```

Arguments

x the result object

Value

the names

dims

Dimension Titles for Prefixes

Description

order does matter, because it defines the order in the dq_report2.

Usage

dims

Format

An object of class character of length 5.

See Also

[util_html_for_var\(\)](#)

[util_html_for_dims\(\)](#)

DISTRIBUTIONS

All available probability distributions for acc_shape_or_scale

Description

- uniform For uniform distribution
- normal For Gaussian distribution
- gamma For a gamma distribution

Usage

DISTRIBUTIONS

Format

An object of class list of length 3.

dq_report	<i>Generate a full DQ report</i>
-----------	----------------------------------

Description

Deprecated

Usage

```
dq_report(...)
```

Arguments

... Deprecated

Value

Deprecated

dq_report2	<i>Generate a full DQ report, v2</i>
------------	--------------------------------------

Description

Generate a full DQ report, v2

Usage

```

dq_report2(
  study_data,
  item_level = "item_level",
  label_col = LABEL,
  meta_data_segment = "segment_level",
  meta_data_dataframe = "dataframe_level",
  meta_data_cross_item = "cross-item_level",
  meta_data_item_computation = "item_computation_level",
  meta_data = item_level,
  meta_data_v2,
  ...,
  dimensions = c("Completeness", "Consistency"),
  cores = list(mode = "socket", logging = FALSE, cpus = util_detect_cores(),
    load.balancing = TRUE),
  specific_args = list(),
  advanced_options = list(),
  author = prep_get_user_name(),
  title = "Data quality report",

```

```

subtitle = as.character(Sys.Date()),
user_info = NULL,
debug_parallel = FALSE,
resp_vars = character(0),
filter_indicator_functions = character(0),
filter_result_slots = c("^Summary", "^Segment", "^DataTypePlotList",
  "^ReportSummaryTable", "^Dataframe", "^Result", "^VariableGroup"),
mode = c("default", "futures", "queue", "parallel"),
mode_args = list(),
notes_from_wrapper = list(),
storr_factory = NULL,
amend = FALSE,
cross_item_level,
`cross-item_level`,
segment_level,
dataframe_level,
item_computation_level,
.internal = rlang::env_inherits(rlang::caller_env(), parent.env(environment()))
)

```

Arguments

study_data [data.frame](#) the data frame that contains the measurements
item_level [data.frame](#) the data frame that contains metadata attributes of study data
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
meta_data_segment
[data.frame](#) – optional: Segment level metadata
meta_data_dataframe
[data.frame](#) – optional: Data frame level metadata
meta_data_cross_item
[data.frame](#) – optional: Cross-item level metadata
meta_data_item_computation
[data.frame](#) optional. computation rules for computed variables.
meta_data [data.frame](#) old name for item_level
meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.
... arguments to be passed to all called indicator functions if applicable.
dimensions [dimensions](#) Vector of dimensions to address in the report. Allowed values in the vector are Completeness, Consistency, and Accuracy. The generated report will only cover the listed data quality dimensions. Accuracy is computational expensive, so this dimension is not enabled by default. Completeness should be included, if Consistency is included, and Consistency should be included, if Accuracy is included to avoid misleading detections of e.g. missing codes as outliers, please refer to the data quality concept for more details. Integrity is always included.

cores	integer number of cpu cores to use or a named list with arguments for <code>parallelMap::parallelStart</code> or <code>NULL</code> , if parallel has already been started by the caller. Can also be a cluster.
specific_args	list named list of arguments specifically for one of the called functions, the of the list elements correspond to the indicator functions whose calls should be modified. The elements are lists of arguments.
advanced_options	list options to set during report computation, see <code>options()</code>
author	character author for the report documents.
title	character optional argument to specify the title for the data quality report
subtitle	character optional argument to specify a subtitle for the data quality report
user_info	list additional info stored with the report, e.g., comments, title, ...
debug_parallel	logical print blocks currently evaluated in parallel
resp_vars	variable list the name of the measurement variables for the report. If missing, all variables will be used. Only item level indicator functions are filtered, so far.
filter_indicator_functions	character regular expressions, only if an indicator function's name matches one of these, it'll be used for the report. If of length zero, no filtering is performed.
filter_result_slots	character regular expressions, only if an indicator function's result's name matches one of these, it'll be used for the report. If of length zero, no filtering is performed.
mode	character work mode for parallel execution. default is "default", the values mean: - default: use queue except cores has been set explicitly - futures: use the future package - queue: use a queue as described in the examples from the callr package by Csárdi and Chang and start sub-processes as workers that evaluate the queue. - parallel: use the cluster from cores to evaluate all calls of indicator functions using the classic R parallel back-ends
mode_args	list of arguments for the selected mode. As of writing this manual, only for the mode queue the argument <code>step</code> is supported, which gives the number of function calls that are run by one worker at a time. the default is 15, which gives on most of the tested systems a good balance between synchronization overhead and idling workers.
notes_from_wrapper	list a list containing notes about changed labels by <code>dq_report_by</code> (otherwise <code>NULL</code>)
storr_factory	function <code>NULL</code> , or a function returning a <code>storr</code> object as back-end for the report's results. If used with <code>cores > 1</code> , the storage must be accessible from all cores and capable of concurrent writing according to <code>storr</code> . Hint: <code>dataquieR</code> currently only supports <code>storr::storr_rds()</code> , officially, while other back-ends may nevertheless work, yet, they are not tested.
amend	logical if there is already data in <code>storr_factory</code> , use it anyways – unsupported, so far!
cross_item_level	data.frame alias for <code>meta_data_cross_item</code>

segment_level [data.frame](#) alias for meta_data_segment
 dataframe_level [data.frame](#) alias for meta_data_dataframe
 item_computation_level [data.frame](#) alias for meta_data_item_computation
 .internal [logical](#) internal use, only.
 'cross-item_level' [data.frame](#) alias for meta_data_cross_item

Details

See [dq_report_by](#) for a way to generate stratified or splitted reports easily.

Value

a [dataquieR_resultset2](#) that can be [printed](#) creating a HTML-report.

See Also

- [as.data.frame.dataquieR_resultset](#)
- [as.list.dataquieR_resultset](#)
- [print.dataquieR_resultset](#)
- [summary.dataquieR_resultset](#)
- [dq_report_by](#)

Examples

```
## Not run:
prep_load_workbook_like_file("inst/extdata/meta_data_v2.xlsx")
meta_data <- prep_get_data_frame("item_level")
meta_data_cross <- prep_get_data_frame("cross-item_level")
x <- dq_report2("study_data", dimensions = NULL, label_col = "LABEL")
xx <- pbapply::pblapply(x, util_eval_to_dataquieR_result, env = environment())
xx <- pbapply::pblapply(tail(x), util_eval_to_dataquieR_result, env = environment())
xx <- parallel
cat(vapply(x, deparse1, FUN.VALUE = character(1)), sep = "\n", file = "all_calls.txt")
rstudioapi::navigateToFile("all_calls.txt")
eval(x$`acc_multivariate_outlier.Blood pressure checks`)

prep_load_workbook_like_file("meta_data_v2")
rules <- tibble::tribble(
  ~resp_vars, ~RULE,
  "BMI", '[BODY_WEIGHT_0]/(([BODY_HEIGHT_0]/100)^2)',
  "R", '[WAIST_CIRC_0]/2/[pi]', # in m^3
  "VOL_EST", '[pi]*([WAIST_CIRC_0]/2/[pi])^2*[BODY_HEIGHT_0] / 1000', # in l
)
prep_load_workbook_like_file("ship_meta_v2")
prep_add_data_frames(computed_items = rules)
r <- dq_report2("ship", dimensions = NULL, label_col = "LABEL")
```



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

```
dq_report_by
```

```
Generate a stratified full DQ report
```

Description

Generate a stratified full DQ report

Usage

```
dq_report_by(
  study_data,
  item_level = "item_level",
  meta_data_segment = "segment_level",
  meta_data_dataframe = "dataframe_level",
  meta_data_cross_item = "cross-item_level",
  meta_data_item_computation = "item_computation_level",
  missing_tables = NULL,
  label_col,
  meta_data_v2,
  segment_column = NULL,
  strata_column = NULL,
  strata_select = NULL,
  selection_type = NULL,
  segment_select = NULL,
  segment_exclude = NULL,
  strata_exclude = NULL,
  subgroup = NULL,
  resp_vars = character(0),
  id_vars = NULL,
  advanced_options = list(),
  storr_factory = NULL,
  amend = FALSE,
  ...,
  output_dir = NULL,
  input_dir = NULL,
  also_print = FALSE,
  disable_plotly = FALSE,
  view = TRUE,
  meta_data = item_level,
  cross_item_level,
  `cross-item_level`,
  segment_level,
  dataframe_level,
  item_computation_level
)
```

Arguments

study_data	data.frame the data frame that contains the measurements: it can be an R object (e.g., <code>bia</code>), a data frame (e.g., <code>"C:/Users/data/bia.dta"</code>), a vector containing data frames files (e.g., <code>c("C:/Users/data/bia.dta", "C:/Users/data/biames.dta")</code>), or it can be left empty and the data frames are provided in the data frame level metadata. If only the file name without path is provided (e.g., <code>"bia.dta"</code>), the file name needs the extension and the path must be provided in the argument <code>input_dir</code> . It can also contain only the file name in case of example data from the package <code>dataquieR</code> (e.g., <code>"study_data"</code> or <code>"ship"</code>)
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data_segment	data.frame – optional: Segment level metadata
meta_data_dataframe	data.frame – optional if <code>study_data</code> is present: Data frame level metadata
meta_data_cross_item	data.frame – optional: Cross-item level metadata
meta_data_item_computation	data.frame – optional: Computed items metadata
missing_tables	character the name of the data frame containing the missing codes, it can be a vector if more than one table is provided. Example: <code>c("missing_table1", "missing_table2")</code>
label_col	variable attribute the name of the column in the metadata containing the labels of the variables
meta_data_v2	character path or file name of the workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code>
segment_column	variable attribute name of a metadata attribute usable to split the report in sections of variables, e.g. all blood-pressure related variables. By default, reports are split by STUDY_SEGMENT if available and no <code>segment_column</code> nor <code>strata_column</code> or <code>subgroup</code> are defined. To create an un-split report please write explicitly the argument <code>'segment_column = NULL'</code>
strata_column	variable name of a study variable to stratify the report by, e.g. the study centers. Both labels and <code>VAR_NAMES</code> are accepted. In case of NAs in the selected variable, a separate report containing the NAs subset will be created
strata_select	character if given, the strata of <code>strata_column</code> are limited to the content of this vector. A character vector or a regular expression can be provided (e.g., <code>"^a.*"</code>). This argument can not be used if no <code>strata_column</code> is provided
selection_type	character optional, can only be specified if a <code>strata_select</code> or <code>strata_exclude</code> is specified. If not present the function try to guess what the user typed as <code>strata_select</code> or <code>strata_exclude</code> . There are 3 options: value indicating that the stratum selected is a value and not a <code>value_label</code> . For example <code>"0"</code> ; <code>v_label</code> indicating that the stratum specified is a label. For example <code>"male"</code> . <code>regex</code> indicating that the user specified strata using a regular expression. For example <code>"^Ber"</code> to select all strata starting with that letters

segment_select	character if given, the levels of segment_column are limited to the content of this vector. A character vector or a regular expression (e.g., ".*_EXAM\$") can be provided. This argument can not be used if no segment_column is provided.
segment_exclude	character optional, can only be specified if a segment_column is specified. The levels of segment_column will not include the content of this argument. A character vector or a regular expression can be provided (e.g., "^STU").
strata_exclude	character optional, can only be specified if a strata_column is specified. The strata of strata_column will not include the content of this argument. A character vector or a regular expression can be provided (e.g., "^STU").
subgroup	character optional, to define subgroups of cases. Rules are to be written as REDCap rules. Only VAR_NAMES are accepted in the rules.
resp_vars	variable the names of the measurement variables, if missing or NULL, all variables will be included
id_vars	variable a vector containing the name/s of the variables containing ids, to be used to merge multiple data frames if provided in study_data and to be add to referred vars
advanced_options	list options to set during report computation, see options()
storr_factory	function NULL, or a function returning a storr object as back-end for the report's results. If used with cores > 1, the storage must be accessible from all cores and capable of concurrent writing according to storr. Hint: dataquieR currently only supports storr::storr_rds(), officially, while other back-ends may nevertheless work, yet, they are not tested.
amend	logical if there is already data in.storr_factory, use it anyways – unsupported, so far!
...	arguments to be passed through to dq_report or dq_report2
output_dir	character if given, the output is not returned but saved in this directory
input_dir	character if given, the study data files that have no path and that are not URL are searched in this directory. Also meta_data_v2 is searched in this directory if no path is provided
also_print	logical if output_dir is not NULL, also create HTML output for each report using print.dataquieR_resultset2() written to the path output_dir
disable_plotly	logical do not use plotly, even if installed
view	logical open the returned report
meta_data	data.frame old name for item_level
cross_item_level	data.frame alias for meta_data_cross_item
segment_level	data.frame alias for meta_data_segment
dataframe_level	data.frame alias for meta_data_dataframe
item_computation_level	data.frame alias for meta_data_item_computation
'cross-item_level'	data.frame alias for meta_data_cross_item

Value

`invisible()`. named [list](#) of named [lists](#) of `dq_report2` reports or, if `output_dir` has been specified, `invisible(NULL)`

See Also

[dq_report](#)

Examples

```
## Not run: # really long-running example.
prep_load_workbook_like_file("meta_data_v2")
rep <- dq_report_by("study_data", label_col =
  LABEL, strata_column = "CENTER_0")
rep <- dq_report_by("study_data",
  label_col = LABEL, strata_column = "CENTER_0",
  segment_column = NULL
)
unlink("/tmp/testRep/", force = TRUE, recursive = TRUE)
dq_report_by("study_data",
  label_col = LABEL, strata_column = "CENTER_0",
  segment_column = STUDY_SEGMENT, output_dir = "/tmp/testRep"
)
unlink("/tmp/testRep/", force = TRUE, recursive = TRUE)
dq_report_by("study_data",
  label_col = LABEL, strata_column = "CENTER_0",
  segment_column = NULL, output_dir = "/tmp/testRep"
)
dq_report_by("study_data",
  label_col = LABEL,
  segment_column = STUDY_SEGMENT, output_dir = "/tmp/testRep"
)
dq_report_by("study_data",
  label_col = LABEL,
  segment_column = STUDY_SEGMENT, output_dir = "/tmp/testRep",
  also_print = TRUE
)
dq_report_by(study_data = "study_data", meta_data_v2 = "meta_data_v2",
  advanced_options = list(dataquieR.study_data_cache_max = 0,
  dataquieR.study_data_cache_metrics = TRUE,
  dataquieR.study_data_cache_metrics_env = environment()),
  cores = NULL, dimensions = "int")
dq_report_by(study_data = "study_data", meta_data_v2 = "meta_data_v2",
  advanced_options = list(dataquieR.study_data_cache_max = 0),
  cores = NULL, dimensions = "int")

## End(Not run)
```

GOLDSTANDARD

Cross-item level metadata attribute name

Description

Defines the measurement variable to be used as a known gold standard. Only one variable can be defined as the gold standard.

Usage

GOLDSTANDARD

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other meta_data_cross: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

html_dependency_clipboard

HTML Dependency for report headers in clipboard

Description

HTML Dependency for report headers in clipboard

Usage

html_dependency_clipboard()

Value

the dependency

html_dependency_dataquieR

HTML Dependency for dataquieR

Description

generate all dependencies used in static dataquieR reports

Usage

```
html_dependency_dataquieR(iframe = FALSE)
```

Arguments

iframe [logical](#)(1) if TRUE, create the dependency used in figure iframes.

Value

the dependency

html_dependency_report_dt

HTML Dependency for report headers in DT::datatable

Description

HTML Dependency for report headers in DT::datatable

Usage

```
html_dependency_report_dt()
```

Value

the dependency

html_dependency_tippy *HTML Dependency for tippy*

Description

HTML Dependency for tippy

Usage

html_dependency_tippy()

Value

the dependency

html_dependency_vert_dt
HTML Dependency for vertical headers in DT::datatable

Description

HTML Dependency for vertical headers in DT::datatable

Usage

html_dependency_vert_dt()

Value

the dependency

int_all_datastructure_dataframe
Wrapper function to check for studies data structure

Description

This function tests for unexpected elements and records, as well as duplicated identifiers and content. The unexpected element record check can be conducted by providing the number of expected records or an additional table with the expected records. It is possible to conduct the checks by study segments or to consider only selected segments.

[Indicator](#)

Usage

```
int_all_datastructure_dataframe(
  meta_data_dataframe = "dataframe_level",
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  dataframe_level
)
```

Arguments

meta_data_dataframe [data.frame](#) the data frame that contains the metadata for the data frame level

item_level [data.frame](#) the data frame that contains metadata attributes of study data

meta_data [data.frame](#) old name for item_level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

dataframe_level [data.frame](#) alias for meta_data_dataframe

Value

a [list](#) with

- DataframeTable: data frame with selected check results, used for the data quality report.

Examples

```
## Not run:
out_dataframe <- int_all_datastructure_dataframe(
  meta_data_dataframe = "meta_data_dataframe",
  meta_data = "ship_meta"
)
md0 <- prep_get_data_frame("ship_meta")
md0
md0$VAR_NAMES
md0$VAR_NAMES[[1]] <- "Id" # is this mismatch reported -- is the data frame
                          # also reported, if nothing is wrong with it
out_dataframe <- int_all_datastructure_dataframe(
  meta_data_dataframe = "meta_data_dataframe",
  meta_data = md0
)

# This is the "normal" procedure for inside pipeline
# but outside this function checktype is exact by default
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "subset_u")
lapply(setNames(nm = prep_get_data_frame("meta_data_dataframe")$DF_NAME),
  int_sts_element_dataframe, meta_data = md0)
md0$VAR_NAMES[[1]] <-
```



```

    "id" # is this mismatch reported -- is the data frame also reported,
        # if nothing is wrong with it
lapply(setNames(nm = prep_get_data_frame("meta_data_dataframe")$DF_NAME),
        int_sts_element_dataframe, meta_data = md0)
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "exact")

## End(Not run)

```

```
int_all_datastructure_segment
```

Wrapper function to check for segment data structure

Description

This function tests for unexpected elements and records, as well as duplicated identifiers and content. The unexpected element record check can be conducted by providing the number of expected records or an additional table with the expected records. It is possible to conduct the checks by study segments or to consider only selected segments.

[Indicator](#)

Usage

```

int_all_datastructure_segment(
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  segment_level,
  meta_data_segment = "segment_level"
)

```

Arguments

study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
segment_level	data.frame alias for meta_data_segment
meta_data_segment	data.frame the data frame that contains the metadata for the segment level, mandatory

Value

a [list](#) with

- SegmentTable: data frame with selected check results, used for the data quality report.

Examples

```
## Not run:
out_segment <- int_all_datastructure_segment(
  meta_data_segment = "meta_data_segment",
  study_data = "ship",
  meta_data = "ship_meta"
)

study_data <- cars
meta_data <- dataquieR::prep_create_meta(VAR_NAMES = c("speedx", "distx"),
  DATA_TYPE = c("integer", "integer"), MISSING_LIST = "|", JUMP_LIST = "|",
  STUDY_SEGMENT = c("Intro", "Ex"))

out_segment <- int_all_datastructure_segment(
  meta_data_segment = "meta_data_segment",
  study_data = study_data,
  meta_data = meta_data
)

## End(Not run)
```

int_datatype_matrix *Check declared data types of metadata in study data*

Description

Checks data types of the study data and for the data type declared in the metadata

[Indicator](#)

Usage

```
int_datatype_matrix(
  resp_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  split_segments = FALSE,
  max_vars_per_plot = 20,
  threshold_value = 0,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

resp_vars	variable the names of the measurement variables, if missing or NULL, all variables will be checked
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
split_segments	logical return one matrix per study segment
max_vars_per_plot	integer from=0. The maximum number of variables per single plot.
threshold_value	numeric from=0 to=100. percentage failing conversions allowed to still classify a study variable convertible.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

This is a preparatory support function that compares study data with associated metadata. A prerequisite of this function is that the no. of columns in the study data complies with the no. of rows in the metadata.

For each study variable, the function searches for its data type declared in static metadata and returns a heatmap like matrix indicating data type mismatches in the study data.

List function.

Value

a list with:

- SummaryTable: data frame containing data quality check for "data type mismatch" (CLS_int_vfe_type, PCT_int_vfe_type). The following categories are possible: categories: "Non-matching datatype", "Non-Matching datatype, convertible", "Matching datatype"
- SummaryData: data frame containing data quality check for "data type mismatch" for a report
- SummaryPlot: [ggplot2::ggplot2](#) heatmap plot, graphical representation of SummaryTable
- DataTypePlotList: [list](#) of plots per (maybe artificial) segment
- ReportSummaryTable: data frame underlying SummaryPlot

int_duplicate_content *Check for duplicated content*

Description

This function tests for duplicated entries in the data set. It is possible to check duplicated entries by study segments or to consider only selected segments.

Indicator

Usage

```
int_duplicate_content(
  level = c("dataframe", "segment"),
  study_data,
  item_level = "item_level",
  label_col,
  meta_data = item_level,
  meta_data_v2,
  ...
)
```

Arguments

level	character a character vector indicating whether the assessment should be conducted at the study level (level = "dataframe") or at the segment level (level = "segment").
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
...	Depending on level, passed to either util_int_duplicate_content_segment or util_int_duplicate_content_dataframe

Value

a **list**. Depending on level, see [util_int_duplicate_content_segment](#) or [util_int_duplicate_content_dataframe](#) for a description of the outputs.

int_duplicate_ids	<i>Check for duplicated IDs</i>
-------------------	---------------------------------

Description

This function tests for duplicated entries in identifiers. It is possible to check duplicated identifiers by study segments or to consider only selected segments.

Indicator

Usage

```
int_duplicate_ids(
  level = c("dataframe", "segment"),
  study_data,
  item_level = "item_level",
  label_col,
  meta_data = item_level,
  meta_data_v2,
  ...
)
```

Arguments

level	character a character vector indicating whether the assessment should be conducted at the study level (level = "dataframe") or at the segment level (level = "segment").
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
...	Depending on level, passed to either util_int_duplicate_ids_segment or util_int_duplicate_ids_dataframe

Value

a **list**. Depending on level, see [util_int_duplicate_ids_segment](#) or [util_int_duplicate_ids_dataframe](#) for a description of the outputs.

int_encoding_errors *Encoding Errors*

Description

Detects errors in the character encoding of string variables

[Indicator](#)

Usage

```
int_encoding_errors(
  resp_vars = NULL,
  study_data,
  label_col,
  meta_data_dataframe = "dataframe_level",
  item_level = "item_level",
  ref_encs,
  meta_data = item_level,
  meta_data_v2,
  dataframe_level
)
```

Arguments

resp_vars	variable the names of the measurement variables, if missing or NULL, all variables will be checked
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data_dataframe	data.frame the data frame that contains the metadata for the data frame level
item_level	data.frame the data frame that contains metadata attributes of study data
ref_encs	reference encodings (names are resp_vars)
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
dataframe_level	data.frame alias for meta_data_dataframe

Details

Strings are stored based on [code tables](#), nowadays, typically as [UTF-8](#). However, other code systems are still in use, so, sometimes, strings from different systems are mixed in the data. This indicator checks for such problems and returns the count of entries per variable, that do not match

the reference coding system, which is estimated from the study data (addition of metadata field is planned).

If not specified in the metadata (columns ENCODING in item- or data-frame- level, the encoding is guessed from the data). Otherwise, it may be any supported encoding as returned by `iconvlist()`.

Value

a list with:

- SummaryTable: [data.frame](#) with information on such problems
- SummaryData: [data.frame](#) human readable version of SummaryTable
- FlaggedStudyData: [data.frame](#) tells for each entry in study data if its encoding is OK. has the same dimensions as `study_data`

int_part_vars_structure

Detect Expected Observations

Description

For each participant, check, if an observation was expected, given the PART_VARS from item-level metadata

Usage

```
int_part_vars_structure(
  label_col,
  study_data,
  item_level = "item_level",
  expected_observations = c("HIERARCHY", "SEGMENT"),
  disclose_problem_paprt_var_data = FALSE,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

`label_col` [character](#) mapping attribute `colnames(study_data)` vs. `meta_data[label_col]`

`study_data` [study_data](#) must have all relevant PART_VARS to avoid false-positives on PART_VARS missing from `study_data`

`item_level` [meta_data](#) must be complete to avoid false positives on non-existing PART_VARS

`expected_observations` [enum](#) HIERARCHY | SEGMENT. How should PART_VARS be handled: - SEGMENT: if PART_VAR is 1, an observation is expected - HIERARCHY: the default, if the PART_VAR is 1 for this variable and also for all PART_VARS of PART_VARS up in the hierarchy, an observation is expected.

disclose_problem_paprt_var_data
 [logical](#) show the problematic data (PART_VAR only)

meta_data [data.frame](#) old name for item_level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Details[Descriptor](#)**Value**

empty list, so far – the function only warns.

int_sts_element_dataframe

Determine missing and/or superfluous data elements

Description

Depends on [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#) option, see there

Usage

```
int_sts_element_dataframe(
  item_level = "item_level",
  meta_data_dataframe = "dataframe_level",
  meta_data = item_level,
  meta_data_v2,
  check_type = getOption("dataquieR.ELEMENT_MISSMATCH_CHECKTYPE",
    dataquieR.ELEMENT_MISSMATCH_CHECKTYPE_default),
  dataframe_level
)
```

Arguments

item_level [data.frame](#) the data frame that contains metadata attributes of study data

meta_data_dataframe
 [data.frame](#) the data frame that contains the metadata for the data frame level

meta_data [data.frame](#) old name for item_level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

check_type [enum](#) none | exact | subset_u | subset_m. See [dataquieR.ELEMENT_MISSMATCH_CHECKTYPE](#)

dataframe_level
 [data.frame](#) alias for meta_data_dataframe

Details[Indicator](#)**Value**[list](#) with names lots:

- DataFrameData: data frame with the unexpected elements check results.
- DataFrameTable: [data.frame](#) table with all errors, used for the data quality report: - PCT_int_sts_element: Percentage of element mismatches - NUM_int_sts_element: Number of element mismatches - resp_vars: affected element names

Examples

```
## Not run:
prep_load_workbook_like_file("~/tmp/df_level_test.xlsx")
meta_data_dataframe <- "dataframe_level"
meta_data <- "item_level"

## End(Not run)
```

int_sts_element_segment

Checks for element set

Description

Depends on dataquieR.ELEMENT_MISSMATCH_CHECKTYPE option, see there – # TODO: Rind out, how to document and link it here using Roxygen.

Usage

```
int_sts_element_segment(
  study_data,
  item_level = "item_level",
  label_col,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

study_data	data.frame the data frame that contains the measurements, mandatory.
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details[Indicator](#)**Value**a [list](#) with

- SegmentData: data frame with the unexpected elements check results. - Segment: name of the corresponding segment, if applicable, ALL otherwise
- SegmentTable: data frame with the unexpected elements check results, used for the data quality report. - Segment: name of the corresponding segment, if applicable, ALL otherwise

Examples

```
## Not run:
study_data <- cars
meta_data <- dataquieR::prep_create_meta(VAR_NAMES = c("speedx", "distx"),
  DATA_TYPE = c("integer", "integer"), MISSING_LIST = "|", JUMP_LIST = "|",
  STUDY_SEGMENT = c("Intro", "Ex"))
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "exact")
int_sts_element_segment(study_data, meta_data)
study_data <- cars
meta_data <- dataquieR::prep_create_meta(VAR_NAMES = c("speedx", "distx"),
  DATA_TYPE = c("integer", "integer"), MISSING_LIST = "|", JUMP_LIST = "|",
  STUDY_SEGMENT = c("Intro", "Intro"))
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "exact")
int_sts_element_segment(study_data, meta_data)
study_data <- cars
meta_data <- dataquieR::prep_create_meta(VAR_NAMES = c("speed", "distx"),
  DATA_TYPE = c("integer", "integer"), MISSING_LIST = "|", JUMP_LIST = "|",
  STUDY_SEGMENT = c("Intro", "Intro"))
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISMATCH_CHECKTYPE = "exact")
int_sts_element_segment(study_data, meta_data)

## End(Not run)
```

`int_unexp_elements`*Check for unexpected data element count*

Description

This function contrasts the expected element number in each study in the metadata with the actual element number in each study data frame.

[Indicator](#)

Usage

```
int_unexp_elements(
  identifier_name_list,
  data_element_count,
  meta_data_dataframe = "dataframe_level",
  meta_data_v2,
  dataframe_level
)
```

Arguments

`identifier_name_list` **character** a character vector indicating the name of each study data frame, mandatory.

`data_element_count` **integer** an integer vector with the number of expected data elements, mandatory.

`meta_data_dataframe` **data.frame** the data frame that contains the metadata for the data frame level

`meta_data_v2` **character** path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`dataframe_level` **data.frame** alias for `meta_data_dataframe`

Value

a **list** with

- **DataframeData**: data frame with the results of the quality check for unexpected data elements
- **DataframeTable**: data frame with selected unexpected data elements check results, used for the data quality report.

`int_unexp_records_dataframe`

Check for unexpected data record count at the data frame level

Description

This function contrasts the expected record number in each study in the metadata with the actual record number in each study data frame.

Indicator

Usage

```
int_unexp_records_dataframe(
  identifier_name_list,
  data_record_count,
  meta_data_dataframe = "dataframe_level",
  meta_data_v2,
  dataframe_level
)
```

Arguments

`identifier_name_list` **character** a character vector indicating the name of each study data frame, mandatory.

`data_record_count` **integer** an integer vector with the number of expected data records per study data frame, mandatory.

`meta_data_dataframe` **data.frame** the data frame that contains the metadata for the data frame level

`meta_data_v2` **character** path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`dataframe_level` **data.frame** alias for `meta_data_dataframe`

Value

a **list** with

- **DataframeData**: data frame with the results of the quality check for unexpected data elements
- **DataframeTable**: data frame with selected unexpected data elements check results, used for the data quality report.

`int_unexp_records_segment`

Check for unexpected data record count within segments

Description

This function contrasts the expected record number in each study segment in the metadata with the actual record number in each segment data frame.

Indicator

Usage

```
int_unexp_records_segment(
  study_segment,
  study_data,
  label_col,
  item_level = "item_level",
  data_record_count,
  meta_data = item_level,
  meta_data_segment = "segment_level",
  meta_data_v2,
  segment_level
)
```

Arguments

`study_segment` **character** a character vector indicating the name of each study data frame, mandatory.

`study_data` **data.frame** the data frame that contains the measurements

`label_col` **variable attribute** the name of the column in the metadata with labels of variables

`item_level` **data.frame** the data frame that contains metadata attributes of study data

`data_record_count` **integer** an integer vector with the number of expected data records, mandatory.

`meta_data` **data.frame** old name for `item_level`

`meta_data_segment` **data.frame** – optional: Segment level metadata

`meta_data_v2` **character** path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`segment_level` **data.frame** alias for `meta_data_segment`

Details

The current implementation does not take into account jump or missing codes, the function is rather based on checking whether NAs are present in the study data

Value

a **list** with

- `SegmentData`: data frame with the results of the quality check for unexpected data elements
- `SegmentTable`: data frame with selected unexpected data elements check results, used for the data quality report.

int_unexp_records_set *Check for unexpected data record set*

Description

This function tests that the identifiers match a provided record set. It is possible to check for unexpected data record sets by study segments or to consider only selected segments.

Indicator

Usage

```
int_unexp_records_set(
  level = c("dataframe", "segment"),
  study_data,
  item_level = "item_level",
  label_col,
  meta_data = item_level,
  meta_data_v2,
  ...
)
```

Arguments

level	character a character vector indicating whether the assessment should be conducted at the study level (level = "dataframe") or at the segment level (level = "segment").
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
...	Depending on level, passed to either util_int_unexp_records_set_segment or util_int_unexp_records_set_dataframe

Value

a **list**. Depending on level, see [util_int_unexp_records_set_segment](#) or [util_int_unexp_records_set_dataframe](#) for a description of the outputs.

meta_data	<i>Data frame with metadata about the study data on variable level</i>
-----------	--

Description

Variable level metadata.

See Also

[further details on variable level metadata.](#)

[meta_data_segment](#)

[meta_data_dataframe](#)

meta_data_cross	<i>Well known columns on the meta_data_cross-item sheet</i>
-----------------	---

Description

Metadata describing groups of variables, e.g., for their multivariate distribution or for defining contradiction rules.

See Also

[check_table](#)

[Online Documentation](#)

Other meta_data_cross: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [util_normalize_cross_item\(\)](#)

meta_data_dataframe	<i>Well known columns on the meta_data_dataframe sheet</i>
---------------------	--

Description

Metadata describing data delivered on one data frame/table sheet, e.g., a full questionnaire, not its items.

meta_data_segment	<i>Well known columns on the meta_data_segment sheet</i>
-------------------	--

Description

Metadata describing study segments, e.g., a full questionnaire, not its items.

MULTIVARIATE_OUTLIER_CHECK	<i>Cross-item level metadata attribute name</i>
----------------------------	---

Description

Select, whether to compute [acc_multivariate_outlier](#).

Usage

MULTIVARIATE_OUTLIER_CHECK

Format

An object of class character of length 1.

Details

You can leave the cell empty, then the depends on the setting of the option [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#). If this column is missing, all this is the same as having all cells empty and [dataquieR.MULTIVARIATE_OUTLIER_CHECK](#) set to "auto".

See also [MULTIVARIATE_OUTLIER_CHECKTYPE](#).

See Also

[meta_data_cross](#)

Other [meta_data_cross](#): [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_it](#)

 MULTIVARIATE_OUTLIER_CHECKTYPE

Cross-item level metadata attribute name

Description

Select, which outlier criteria to compute, see [acc_multivariate_outlier](#).

Usage

MULTIVARIATE_OUTLIER_CHECKTYPE

Format

An object of class character of length 1.

Details

You can leave the cell empty, then, all checks will apply. If you enter a set of methods, the maximum for [N_RULES](#) changes. See also [UNIVARIATE_OUTLIER_CHECKTYPE](#).

See Also

[meta_data_cross](#)

Other [meta_data_cross](#): [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [N_RULES](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

 nres

return the number of result slots in a report

Description

return the number of result slots in a report

Usage

nres(x)

Arguments

x the dataquieR report (v2.0)

Value

the number of used result slots

N_RULES	<i>Cross-item and item level metadata attribute name</i>
---------	--

Description

Select, how many violated outlier criteria make an observation an outlier, see [acc_multivariate_outlier](#).

Usage

N_RULES

Format

An object of class character of length 1.

Details

You can leave the cell empty, then, all applied checks must deem an observation an outlier to have it flagged. See [UNIVARIATE_OUTLIER_CHECKTYPE](#) and [MULTIVARIATE_OUTLIER_CHECKTYPE](#) for the selected outlier criteria.

See Also

[meta_data_cross](#)

[meta_data](#)

Other [meta_data_cross](#): [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [REL_VAL](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

pipeline_recursive_result

Convert a pipeline result data frame to named encapsulated lists

Description

Deprecated

Usage

pipeline_recursive_result(...)

Arguments

... Deprecated

Value

Deprecated

pipeline_vectorized	<i>Call (nearly) one "Accuracy" function with many parameterizations at once automatically</i>
---------------------	--

Description

Deprecated

Usage

```
pipeline_vectorized(...)
```

Arguments

... Deprecated

Value

Deprecated

plot.dataquieR_summary	<i>Plot a dataquieR summary</i>
------------------------	---------------------------------

Description

Plot a dataquieR summary

Usage

```
## S3 method for class 'dataquieR_summary'
plot(x, y, ..., filter, dont_plot = FALSE, stratify_by)
```

Arguments

x	the dataquieR summary, see summary() and dq_report2()
y	not yet used
...	not yet used
filter	if given, this filters the summary, e.g., filter = call_names == "com_qualified_item_missingness"
dont_plot	suppress the actual plotting, just return a printable object derived from x
stratify_by	column to stratify the summary, may be one string.

Value

invisible html object

```
prep_acc_distributions_with_ecdf
```

Utility function to plot a combined figure for distribution checks

Description

Data quality indicator checks "Unexpected location" with histograms and plots of empirical cumulative distributions for the subgroups.

Usage

```
prep_acc_distributions_with_ecdf(
  resp_vars = NULL,
  group_vars = NULL,
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  n_group_max = getOption("dataquieR.max_group_var_levels_in_plot",
    dataquieR.max_group_var_levels_in_plot_default),
  n_obs_per_group_min = getOption("dataquieR.min_obs_per_group_var_in_plot",
    dataquieR.min_obs_per_group_var_in_plot_default)
)
```

Arguments

resp_vars	variable list the name of the measurement variable
group_vars	variable list the name of the observer, device or reader variable
study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
n_group_max	maximum number of categories to be displayed individually for the grouping variable (group_vars, devices / examiners)
n_obs_per_group_min	minimum number of data points per group to create a graph for an individual category of the group_vars variable

Value

A SummaryPlot.

```
prep_add_cause_label_df
```

Convert missing codes in metadata format v1.0 and a missing-cause-table to v2.0 missing list / jump list assignments

Description

The function has two working modes. If `replace_meta_data` is TRUE, by default, if `cause_label_df` contains a column named `resp_vars`, then the missing/jump codes in `meta_data[, c(MISSING_CODES, JUMP_CODES)]` will be overwritten, otherwise, it will be labeled using the `cause_label_df`.

Usage

```
prep_add_cause_label_df(
  item_level = "item_level",
  cause_label_df,
  label_col = VAR_NAMES,
  assume_consistent_codes = TRUE,
  replace_meta_data = ("resp_vars" %in% colnames(cause_label_df)),
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

`item_level` [data.frame](#) the data frame that contains metadata attributes of study data

`cause_label_df` [data.frame](#) missing code table. If missing codes have labels the respective data frame can be specified here, see [cause_label_df](#)

`label_col` [variable attribute](#) the name of the column in the metadata with labels of variables

`assume_consistent_codes` [logical](#) if TRUE and no labels are given and the same missing/jump code is used for more than one variable, the labels assigned for this code will be the same for all variables.

`replace_meta_data` [logical](#) if TRUE, ignore existing missing codes and jump codes and replace them with data from the `cause_label_df`. Otherwise, copy the labels from `cause_label_df` to the existing code columns.

`meta_data` [data.frame](#) old name for `item_level`

`meta_data_v2` [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

Details

If a column `resp_vars` exists, then rows with a value in `resp_vars` will only be used for the corresponding variable.

Value

`data.frame` updated metadata including all the code labels in missing/jump lists

See Also

[prep_extract_cause_label_df](#)

prep_add_computed_variables

Insert missing codes for NAs based on rules

Description

Insert missing codes for NAs based on rules

Usage

```
prep_add_computed_variables(
  study_data,
  meta_data,
  label_col,
  rules,
  use_value_labels
)
```

Arguments

<code>study_data</code>	data.frame the data frame that contains the measurements
<code>meta_data</code>	data.frame the data frame that contains metadata attributes of study data
<code>label_col</code>	variable attribute the name of the column in the metadata with labels of variables
<code>rules</code>	data.frame with the columns: <ul style="list-style-type: none"> • <code>VAR_NAMES</code>: VAR_NAMES of the variable to compute • <code>RULE</code>: A rule in REDcap style (see, e.g., REDcap help, REDcap how-to), and REDcap branching logic that defines, how to compute the new values
<code>use_value_labels</code>	logical In rules for factors, use the value labels, not the codes. Defaults to TRUE, if any <code>VALUE_LABELS</code> are given in the metadata.

Value

a list with the entry:

- ModifiedStudyData: Study data with the new variables

Examples

```
## Not run:
study_data <- prep_get_data_frame("ship")
prep_load_workbook_like_file("ship_meta_v2")
meta_data <- prep_get_data_frame("item_level")
rules <- tibble::tribble(
  ~VAR_NAMES, ~RULE,
  "BMI", '[BODY_WEIGHT_0]/(([BODY_HEIGHT_0]/100)^2)',
  "R", '[WAIST_CIRC_0]/2/[pi]', # in m^3
  "VOL_EST", '[pi]*([WAIST_CIRC_0]/2/[pi])^2*[BODY_HEIGHT_0] / 1000', # in l
)
r <- prep_add_computed_variables(study_data, meta_data,
  label_col = "LABEL", rules, use_value_labels = FALSE)

## End(Not run)
```

prep_add_data_frames *Add data frames to the pre-loaded / cache data frame environment*

Description

These can be referred to by their names, then, wherever dataquieR expects a [data.frame](#) – just pass a character instead. If this character is not found, dataquieR would additionally look for files with the name and for URLs. You can also refer to specific sheets of a workbook or specific object from an RData by appending a pipe symbol and its name. A second pipe symbol allows to extract certain columns from such sheets (but they will remain data frames).

Usage

```
prep_add_data_frames(..., data_frame_list = list())
```

Arguments

... data frames, if passed with names, these will be the names of these tables in the data frame environment. If not, then the names in the calling environment will be used.

data_frame_list a named list with data frames. Also these will be added and names will be handled as for the ... argument.

Value

[data.frame](#) invisible(the cache environment)

See Also[prep_load_workbook_like_file](#)[prep_get_data_frame](#)

Other data-frame-cache: [prep_get_data_frame\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_purge_data_frame_cache\(\)](#), [prep_remove_from_cache\(\)](#)

prep_add_missing_codes

Insert missing codes for NAs based on rules

Description

Insert missing codes for NAs based on rules

Usage

```
prep_add_missing_codes(
  resp_vars,
  study_data,
  meta_data_v2,
  item_level = "item_level",
  label_col,
  rules,
  use_value_labels,
  overwrite = FALSE,
  meta_data = item_level
)
```

Arguments

resp_vars	variable list the name of the measurement variables to be modified, all from rules, if omitted
study_data	data.frame the data frame that contains the measurements
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
rules	data.frame with the columns: <ul style="list-style-type: none"> • resp_vars: Variable, whose NA-values should be replaced by jump codes • CODE_CLASS: Either MISSING or JUMP: Is the currently described case an expected missing value (JUMP) or not (MISSING) • CODE_VALUE: The jump code or missing code • CODE_LABEL: A label describing the reason for the missing value

- **RULE:** A rule in REDcap style (see, e.g., [REDcap help](#), [REDcap how-to](#)), and [REDcap branching logic](#) that describes cases for the missing
- use_value_labels [logical](#) In rules for factors, use the value labels, not the codes. Defaults to TRUE, if any VALUE_LABELS are given in the metadata.
- overwrite [logical](#) Also insert missing codes, if the values are not NA
- meta_data [data.frame](#) old name for item_level attributes of study data

Value

a list with the entries:

- ModifiedStudyData: Study data with NAs replaced by the CODE_VALUE
- ModifiedMetaData: Metadata having the new codes amended in the columns JUMP_LIST or MISSING_LIST, respectively

prep_add_to_meta *Support function to augment metadata during data quality reporting*

Description

adds an annotation to static metadata

Usage

```
prep_add_to_meta(
  VAR_NAMES,
  DATA_TYPE,
  LABEL,
  VALUE_LABELS,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  ...
)
```

Arguments

- VAR_NAMES [character](#) Names of the Variables to add
- DATA_TYPE [character](#) Data type for the added variables
- LABEL [character](#) Labels for these variables
- VALUE_LABELS [character](#) Value labels for the values of the variables as usually pipe separated and assigned with =: 1 = male | 2 = female
- item_level [data.frame](#) the metadata to extend
- meta_data [data.frame](#) old name for item_level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

... Further defined variable attributes, see [prep_create_meta](#)

Details

Add metadata e.g. of transformed/new variable This function is not yet considered stable, but we already export it, because it could help. Therefore, we have some inconsistencies in the formal still.

Value

a data frame with amended metadata.

prep_apply_coding	<i>Re-Code labels with their respective codes according to the meta_data</i>
-------------------	--

Description

Re-Code labels with their respective codes according to the meta_data

Usage

```
prep_apply_coding(
  study_data,
  meta_data_v2,
  item_level = "item_level",
  meta_data = item_level
)
```

Arguments

study_data [data.frame](#) the data frame that contains the measurements

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

item_level [data.frame](#) the data frame that contains metadata attributes of study data

meta_data [data.frame](#) old name for item_level

Value

[data.frame](#) modified study data with labels replaced by the codes

prep_check_for_dataquieR_updates
Check for package updates

Description

Check for package updates

Usage

```
prep_check_for_dataquieR_updates(  
  beta = FALSE,  
  deps = TRUE,  
  ask = interactive()  
)
```

Arguments

beta **logical** check for beta version too
deps **logical** check for missing (optional) dependencies
ask **logical** ask for updates

Value

invisible(NULL)

prep_check_meta_data_dataframe
Verify and normalize metadata on data frame level

Description

if possible, mismatching data types are converted ("true" becomes TRUE)

Usage

```
prep_check_meta_data_dataframe(  
  meta_data_dataframe = "dataframe_level",  
  meta_data_v2,  
  dataframe_level  
)
```

Arguments

meta_data_dataframe [data.frame](#) data frame or path/url of a metadata sheet for the data frame level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

dataframe_level [data.frame](#) alias for meta_data_dataframe

Details

missing columns are added, filled with NA, if this is valid, i.e., n.a. for DF_NAME as the key column

Value

standardized metadata sheet as data frame

Examples

```
## Not run:
mds <- prep_check_meta_data_dataframe("ship_meta_dataframe|dataframe_level") # also converts
print(mds)
prep_check_meta_data_dataframe(mds)
mds1 <- mds
mds1$DF_RECORD_COUNT <- NULL
print(prepare_check_meta_data_dataframe(mds1)) # fixes the missing column by NAs
mds1 <- mds
mds1$DF_UNIQUE_ROWS[[2]] <- "xxx" # not convertible
# print(prepare_check_meta_data_dataframe(mds1)) # fail
mds1 <- mds
mds1$DF_UNIQUE_ID[[2]] <- 12
# print(prepare_check_meta_data_dataframe(mds1)) # fail

## End(Not run)
```

```
prep_check_meta_data_segment
```

Verify and normalize metadata on segment level

Description

if possible, mismatching data types are converted ("true" becomes TRUE)

Usage

```
prep_check_meta_data_segment(
  meta_data_segment = "segment_level",
  meta_data_v2,
  segment_level
)
```

Arguments

meta_data_segment [data.frame](#) data frame or path/url of a metadata sheet for the segment level

meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

segment_level [data.frame](#) alias for meta_data_segment

Details

missing columns are added, filled with NA, if this is valid, i.e., n.a. for STUDY_SEGMENT as the key column

Value

standardized metadata sheet as data frame

Examples

```
## Not run:
mds <- prep_check_meta_data_segment("ship_meta_v2|segment_level") # also converts
print(mds)
prep_check_meta_data_segment(mds)
mds1 <- mds
mds1$SEGMENT_RECORD_COUNT <- NULL
print(prepare_check_meta_data_segment(mds1)) # fixes the missing column by NAs
mds1 <- mds
mds1$SEGMENT_UNIQUE_ROWS[[2]] <- "xxx" # not convertible
# print(prepare_check_meta_data_segment(mds1)) # fail

## End(Not run)
```

prep_check_meta_names *Checks the validity of metadata w.r.t. the provided column names*

Description

This function verifies, if a data frame complies to metadata conventions and provides a given richness of meta information as specified by level.

Usage

```
prep_check_meta_names(
  item_level = "item_level",
  level,
  character.only = FALSE,
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

item_level	data.frame the data frame that contains metadata attributes of study data
level	enum level of requirement (see also VARATT_REQUIRE_LEVELS). set to NULL to deactivate the check of richness.
character.only	logical a logical indicating whether level can be assumed to be character strings.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

Note, that only the given level is checked despite, levels are somehow hierarchical.

Value

a logical with:

- invisible(TRUE). In case of problems with the metadata, a condition is raised (stop()).

Examples

```
## Not run:
prep_check_meta_names(data.frame(VAR_NAMES = 1, DATA_TYPE = 2,
                                MISSING_LIST = 3))

prep_check_meta_names(
  data.frame(
    VAR_NAMES = 1, DATA_TYPE = 2, MISSING_LIST = 3,
    LABEL = "LABEL", VALUE_LABELS = "VALUE_LABELS",
    JUMP_LIST = "JUMP_LIST", HARD_LIMITS = "HARD_LIMITS",
    GROUP_VAR_OBSERVER = "GROUP_VAR_OBSERVER",
    GROUP_VAR_DEVICE = "GROUP_VAR_DEVICE",
    TIME_VAR = "TIME_VAR",
    PART_VAR = "PART_VAR",
    STUDY_SEGMENT = "STUDY_SEGMENT",
    LOCATION_RANGE = "LOCATION_RANGE",
    LOCATION_METRIC = "LOCATION_METRIC",
    PROPORTION_RANGE = "PROPORTION_RANGE",
    MISSING_LIST_TABLE = "MISSING_LIST_TABLE",
    CO_VARS = "CO_VARS",
    LONG_LABEL = "LONG_LABEL"
  ),
  RECOMMENDED
)

prep_check_meta_names(
  data.frame(
    VAR_NAMES = 1, DATA_TYPE = 2, MISSING_LIST = 3,
    LABEL = "LABEL", VALUE_LABELS = "VALUE_LABELS",
```

```

    JUMP_LIST = "JUMP_LIST", HARD_LIMITS = "HARD_LIMITS",
    GROUP_VAR_OBSERVER = "GROUP_VAR_OBSERVER",
    GROUP_VAR_DEVICE = "GROUP_VAR_DEVICE",
    TIME_VAR = "TIME_VAR",
    PART_VAR = "PART_VAR",
    STUDY_SEGMENT = "STUDY_SEGMENT",
    LOCATION_RANGE = "LOCATION_RANGE",
    LOCATION_METRIC = "LOCATION_METRIC",
    PROPORTION_RANGE = "PROPORTION_RANGE",
    DETECTION_LIMITS = "DETECTION_LIMITS", SOFT_LIMITS = "SOFT_LIMITS",
    CONTRADICTIONS = "CONTRADICTIONS", DISTRIBUTION = "DISTRIBUTION",
    DECIMALS = "DECIMALS", VARIABLE_ROLE = "VARIABLE_ROLE",
    DATA_ENTRY_TYPE = "DATA_ENTRY_TYPE",
    CO_VARS = "CO_VARS",
    END_DIGIT_CHECK = "END_DIGIT_CHECK",
    VARIABLE_ORDER = "VARIABLE_ORDER", LONG_LABEL =
      "LONG_LABEL", recode = "recode",
    MISSING_LIST_TABLE = "MISSING_LIST_TABLE"
  ),
  OPTIONAL
)

# Next one will fail
try(
  prep_check_meta_names(data.frame(VAR_NAMES = 1, DATA_TYPE = 2,
    MISSING_LIST = 3), TECHNICAL)
)

## End(Not run)

```

```
prep_clean_labels      Support function to scan variable labels for applicability
```

Description

Adjust labels in `meta_data` to be valid variable names in formulas for diverse r functions, such as `glm` or `lme4::lmer`.

Usage

```

prep_clean_labels(
  label_col,
  item_level = "item_level",
  no_dups = FALSE,
  meta_data = item_level,
  meta_data_v2
)

```

Arguments

label_col	character label attribute to adjust or character vector to adjust, depending on meta_data argument is given or missing.
item_level	data.frame metadata data frame: If label_col is a label attribute to adjust, this is the metadata table to process on. If missing, label_col must be a character vector with values to adjust.
no_dups	logical disallow duplicates in input or output vectors of the function, then, prep_clean_labels would call stop() on duplicated labels.
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Details

Hint: The following is still true, but the functions should be capable of doing potentially needed fixes on-the-fly automatically, so likely you will not need this function any more.

Currently, labels as given by label_col arguments in the most functions are directly used in formula, so that they become natural part of the outputs, but different models expect differently strict syntax for such formulas, especially for valid variable names. prep_clean_labels removes all potentially inadmissible characters from variable names (no guarantee, that some exotic model still rejects the names, but minimizing the number of exotic characters). However, variable names are modified, may become unreadable or indistinguishable from other variable names. For the latter case, a stop call is possible, controlled by the no_dups argument.

A warning is emitted, if modifications were necessary.

Value

a data.frame with:

- if meta_data is set, a list with:
 - modified meta_data[, label_col] column
- if meta_data is not set, adjusted labels that then were directly given in label_col

Examples

```
## Not run:
meta_data1 <- data.frame(
  LABEL =
    c(
      "syst. Blood pressure (mmHg) 1",
      "1st heart frequency in MHz",
      "body surface (\u33A1)"
    )
)
print(meta_data1)
print(prepare_clean_labels(meta_data1$LABEL))
```



```
meta_data1 <- prep_clean_labels("LABEL", meta_data1)
print(meta_data1)

## End(Not run)
```

prep_combine_report_summaries
Combine two report summaries

Description

Combine two report summaries

Usage

```
prep_combine_report_summaries(..., summaries_list, amend_segment_names = FALSE)
```

Arguments

... objects returned by [prep_extract_summary](#)
summaries_list if given, [list](#) of objects returned by [prep_extract_summary](#)
amend_segment_names [logical](#) use names of the summaries_list and argument names as segment prefixes

Value

combined summaries

See Also

Other summary_functions: [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_extract_summary.dataquieR_resultset2\(\)](#), [prep_render_pie_chart_from_summaryclasses_ggplot2\(\)](#), [prep_render_pie_chart_from_summaryclasses_plot1](#), [prep_summary_to_classes\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

```
prep_compare_meta_with_study
      Verify item-level metadata
```

Description

are the provided item-level meta_data plausible given study_data?

Usage

```
prep_compare_meta_with_study(
  study_data,
  label_col,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2
)
```

Arguments

study_data	data.frame the data frame that contains the measurements
label_col	variable attribute the name of the column in the metadata with labels of variables
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.

Value

an [invisible\(\)](#) list with the entries.

- pred [data.frame](#) metadata predicted from study_data, reduced to such metadata also available in the provided metadata
- prov [data.frame](#) provided metadata, reduced to such metadata also available in the provided study_data
- ml_error [character](#) VAR_NAMES of variables with potentially wrong MISSING_LIST
- sl_error [character](#) VAR_NAMES of variables with potentially wrong SCALE_LEVEL
- dt_error [character](#) VAR_NAMES of variables with potentially wrong DATA_TYPE

```
prep_create_meta      Support function to create data.frames of metadata
```

Description

Create a metadata data frame and map names. Generally, this function only creates a [data.frame](#), but using this constructor instead of calling `data.frame(..., stringsAsFactors = FALSE)`, it becomes possible, to adapt the metadata [data.frame](#) in later developments, e.g. if we decide to use classes for the metadata, or if certain standard names of variable attributes change. Also, a validity check is possible to implement here.

Usage

```
prep_create_meta(..., stringsAsFactors = FALSE, level, character.only = FALSE)
```

Arguments

`...` named column vectors, names will be mapped using [WELL_KNOWN_META_VARIABLE_NAMES](#), if included in [WELL_KNOWN_META_VARIABLE_NAMES](#) can also be a data frame, then its column names will be mapped using [WELL_KNOWN_META_VARIABLE_NAMES](#)

`stringsAsFactors` [logical](#) if the argument is a list of vectors, a data frame will be created. In this case, `stringsAsFactors` controls, whether characters will be auto-converted to Factors, which defaults here always to false independent from the [default.stringsAsFactors](#).

`level` [enum](#) level of requirement (see also [VARATT_REQUIRE_LEVELS](#)) set to NULL, if not a complete metadata frame is created.

`character.only` [logical](#) a logical indicating whether level can be assumed to be character strings.

Details

For now, this calls [data.frame](#), but it already renames variable attributes, if they have a different name assigned in [WELL_KNOWN_META_VARIABLE_NAMES](#), e.g. `WELL_KNOWN_META_VARIABLE_NAMES$RECODE` maps to `recode` in lower case.

NB: `dataquieR` exports all names from `WELL_KNOWN_META_VARIABLE_NAME` as symbols, so `RECODE` also contains "recode".

Value

a data frame with:

- metadata attribute names mapped and
- metadata checked using [prep_check_meta_names](#) and do some more verification about conventions, such as check for valid intervals in limits)

See Also

[WELL_KNOWN_META_VARIABLE_NAMES](#)

prep_create_meta_data_file

Instantiate a new metadata file

Description

Instantiate a new metadata file

Usage

```
prep_create_meta_data_file(  
  file_name,  
  study_data,  
  open = TRUE,  
  overwrite = FALSE  
)
```

Arguments

file_name **character** file path to write to
study_data **data.frame** optional, study data to guess metadata from
open **logical** open the file after creation
overwrite **logical** overwrite file, if exists

Value

invisible(NULL)

prep_create_storr_factory

Create a factory function for storr objects for backing a [dataquieR_resultset2](#)

Description

Create a factory function for storr objects for backing a [dataquieR_resultset2](#)

Usage

```
prep_create_storr_factory(db_dir = tempfile(), namespace = "objects")
```

Arguments

`db_dir` **character** path to the directory for the back-end, if one is created on the fly.
`namespace` **character** namespace for the report, so that one back-end can back several reports
the returned function will try to create a `storr` object using a temporary folder or the folder in `db_dir`, if specified. The database will either be the `storr_rds`.

Value

`storr` object or `NULL`, if package `storr` is not available

`prep_datatype_from_data`*Get data types from data*

Description

Get data types from data

Usage

```
prep_datatype_from_data(  
  resp_vars = colnames(study_data),  
  study_data,  
  .dont_cast_off_cols = FALSE  
)
```

Arguments

`resp_vars` **variable** names of the variables to fetch the data type from the data
`study_data` **data.frame** the data frame that contains the measurements Hint: Only data frames supported, no URL or file names.
`.dont_cast_off_cols` **logical** internal use, only

Value

vector of data types

Examples

```
## Not run:  
dataquieR::prep_datatype_from_data(cars)  
  
## End(Not run)
```

```
prep_deparse_assignments
```

Convert two vectors from a code-value-table to a key-value list

Description

Convert two vectors from a code-value-table to a key-value list

Usage

```
prep_deparse_assignments(
  codes,
  labels = codes,
  split_char = SPLIT_CHAR,
  mode = c("numeric_codes", "string_codes")
)
```

Arguments

codes	codes, numeric or dates (as default, but string codes can be enabled using the option 'mode', see below)
labels	character labels, same length as codes
split_char	character split character character to split code assignments
mode	character one of two options to insist on numeric or datetime codes (default) or to allow for string codes

Value

a vector with assignment strings for each row of `cbind(codes, labels)`

```
prep_dq_data_type_of Get the dataquieR DATA_TYPE of x
```

Description

Get the dataquieR DATA_TYPE of x

Usage

```
prep_dq_data_type_of(x)
```

Arguments

x	object to define the dataquieR data type of
---	---

Value

the dataquieR data type as listed in DATA_TYPES

See Also

[DATA_TYPES_OF_R_TYPE](#)

prep_expand_codes	<i>Expand code labels across variables</i>
-------------------	--

Description

Code labels are copied from other variables, if the code is the same and the label is set only for some variables

Usage

```
prep_expand_codes(
  item_level = "item_level",
  suppressWarnings = FALSE,
  mix_jumps_and_missings = FALSE,
  meta_data_v2,
  meta_data = item_level
)
```

Arguments

`item_level` [data.frame](#) the data frame that contains metadata attributes of study data

`suppressWarnings` [logical](#) show warnings, if labels are expanded

`mix_jumps_and_missings` [logical](#) ignore the class of the codes for label expansion, i.e., use missing code labels as jump code labels, if the values are the same.

`meta_data_v2` [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`meta_data` [data.frame](#) old name for `item_level`

Value

[data.frame](#) an updated metadata data frame.

Examples

```
## Not run:
meta_data <- prep_get_data_frame("meta_data")
meta_data$JUMP_LIST[meta_data$VAR_NAMES == "v00003"] <- "99980 = NOOP"
md <- prep_expand_codes(meta_data)
md$JUMP_LIST
md$MISSING_LIST
md <- prep_expand_codes(meta_data, mix_jumps_and_missings = TRUE)
md$JUMP_LIST
md$MISSING_LIST
meta_data <- prep_get_data_frame("meta_data")
meta_data$MISSING_LIST[meta_data$VAR_NAMES == "v00003"] <- "99980 = NOOP"
md <- prep_expand_codes(meta_data)
md$JUMP_LIST
md$MISSING_LIST

## End(Not run)
```

```
prep_extract_cause_label_df
```

Extract all missing/jump codes from metadata and export a cause-label-data-frame

Description

Extract all missing/jump codes from metadata and export a cause-label-data-frame

Usage

```
prep_extract_cause_label_df(
  item_level = "item_level",
  label_col = VAR_NAMES,
  meta_data_v2,
  meta_data = item_level
)
```

Arguments

item_level	data.frame the data frame that contains metadata attributes of study data
label_col	variable attribute the name of the column in the metadata with labels of variables
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
meta_data	data.frame old name for item_level

Value

`list` with the entries

- `meta_data` `data.frame` a data frame that contains updated metadata – you still need to add a column `MISSING_LIST_TABLE` and add the `cause_label_df` as such to the metadata cache using `prep_add_data_frames()`, manually.
- `cause_label_df` `data.frame` missing code table. If missing codes have labels the respective data frame are specified here, see `cause_label_df`.

See Also

`prep_add_cause_label_df`

`prep_extract_classes_by_functions`

Extract old function based summary from data quality results

Description

Extract old function based summary from data quality results

Usage

```
prep_extract_classes_by_functions(r)
```

Arguments

`r` `dq_report2`

Value

`data.frame` long format, compatible with `prep_summary_to_classes()`

See Also

Other `summary_functions`: `prep_combine_report_summaries()`, `prep_extract_summary()`, `prep_extract_summary.d`, `prep_extract_summary.dataquieR_resultset2()`, `prep_render_pie_chart_from_summaryclasses_ggplot2()`, `prep_render_pie_chart_from_summaryclasses_plotly()`, `prep_summary_to_classes()`, `util_as_cat()`, `util_as_integer_cat()`, `util_extract_indicator_metrics()`, `util_get_category_for_result()`, `util_get_colors()`, `util_get_labels_grading_class()`, `util_get_message_for_result()`, `util_get_rule_sets()`, `util_get_ruleset_formats()`, `util_get_thresholds()`, `util_html_table()`, `util_sort_by_order()`

```
prep_extract_summary Extract summary from data quality results
```

Description

Generic function, currently supports [dq_report2](#) and [dataquieR_result](#)

Usage

```
prep_extract_summary(r, ...)
```

Arguments

`r` [dq_report2](#) or [dataquieR_result](#) object
`...` further arguments, maybe needed for some implementations

Value

[list](#) with two slots `Data` and `Table` with [data.frames](#) featuring all metrics columns from the report or result in `x`, the [STUDY_SEGMENT](#) and the [VAR_NAMES](#). In case of `Data`, the columns are formatted nicely but still with the standardized column names – use [util_translate_indicator_metrics\(\)](#) to rename them nicely. In case of `Table`, just as they are.

See Also

Other `summary_functions`: [prep_combine_report_summaries\(\)](#), [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_extract_summary.dataquieR_resultset2\(\)](#), [prep_render_pie_chart_from_summaryclasses_ggplot2\(\)](#), [prep_render_pie_chart_from_summaryclasses_plot1\(\)](#), [prep_summary_to_classes\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

```
prep_extract_summary.dataquieR_result  

Extract report summary from reports
```

Description

Extract report summary from reports

Usage

```
## S3 method for class 'dataquieR_result'  

prep_extract_summary(r, ...)
```

Arguments

`r` [dataquieR_result](#) a result from `adq_report2` report
`...` not used

Value

`list` with two slots `Data` and `Table` with `data.frames` featuring all metrics columns from the report `r`, the `STUDY_SEGMENT` and the `VAR_NAMES`. In case of `Data`, the columns are formatted nicely but still with the standardized column names – use `util_translate_indicator_metrics()` to rename them nicely. In case of `Table`, just as they are.

See Also

`prep_combine_report_summaries()`

Other `summary_functions`: `prep_combine_report_summaries()`, `prep_extract_classes_by_functions()`, `prep_extract_summary()`, `prep_extract_summary.dataquieR_resultset2()`, `prep_render_pie_chart_from_summary()`, `prep_render_pie_chart_from_summaryclasses_plotly()`, `prep_summary_to_classes()`, `util_as_cat()`, `util_as_integer_cat()`, `util_extract_indicator_metrics()`, `util_get_category_for_result()`, `util_get_colors()`, `util_get_labels_grading_class()`, `util_get_message_for_result()`, `util_get_rule_sets()`, `util_get_ruleset_formats()`, `util_get_thresholds()`, `util_html_table()`, `util_sort_by_order()`

```
prep_extract_summary.dataquieR_resultset2
```

Extract report summary from reports

Description

Extract report summary from reports

Usage

```
## S3 method for class 'dataquieR_resultset2'
prep_extract_summary(r, ...)
```

Arguments

`r` [dq_report2](#) a `dq_report2` report
`...` not used

Value

`list` with two slots `Data` and `Table` with `data.frames` featuring all metrics columns from the report `r`, the `STUDY_SEGMENT` and the `VAR_NAMES`. In case of `Data`, the columns are formatted nicely but still with the standardized column names – use `util_translate_indicator_metrics()` to rename them nicely. In case of `Table`, just as they are.

See Also

[prep_combine_report_summaries\(\)](#)

Other `summary_functions`: [prep_combine_report_summaries\(\)](#), [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_render_pie_chart_from_summaryclasses\(\)](#), [prep_render_pie_chart_from_summaryclasses_plotly\(\)](#), [prep_summary_to_classes\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

prep_get_data_frame *Read data from files/URLs*

Description

`data_frame_name` can be a file path or an URL you can append a pipe and a sheet name for Excel files or object name e.g. for RData files. Numbers may also work. All file formats supported by your rio installation will work.

Usage

```
prep_get_data_frame(
  data_frame_name,
  .data_frame_list = .dataframe_environment(),
  keep_types = FALSE,
  column_names_only = FALSE
)
```

Arguments

`data_frame_name` **character** name of the data frame to read, see details

`.data_frame_list` **environment** cache for loaded data frames

`keep_types` **logical** keep types as possibly defined in a file, if the data frame is loaded from one. set TRUE for study data.

`column_names_only` **logical** if TRUE imports only headers (column names) of the data frame and no content (an empty data frame)

Details

The data frames will be cached automatically, you can define an alternative environment for this using the argument `.data_frame_list`, and you can purge the cache using [prep_purge_data_frame_cache](#).

Use [prep_add_data_frames](#) to manually add data frames to the cache, e.g., if you have loaded them from more complex sources, before.

Value

`data.frame` a data frame

See Also

[prep_add_data_frames](#)

[prep_load_workbook_like_file](#)

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_purge_data_frame_cache\(\)](#), [prep_remove_from_cache\(\)](#)

Examples

```
## Not run:
bl <- as.factor(prepare_get_data_frame(
  paste0("https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus",
    "/Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=",
    "publicationFile|COVID_Todesfaelle_BL|Bundesland"))[[1]])

n <- as.numeric(prepare_get_data_frame(paste0(
  "https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus/",
  "Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=",
  "publicationFile|COVID_Todesfaelle_BL|Anzahl_verstorbene",
  " COVID-19 Falle"))[[1]])
plot(bl, n)
# Working names would be to date (2022-10-21), e.g.:
#
# https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus/ \
#   Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=publicationFile
# https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus/ \
#   Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=publicationFile|2
# https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus/ \
#   Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=publicationFile|name
# study_data
# ship
# meta_data
# ship_meta
#
prepare_get_data_frame("meta_data | meta_data")

## End(Not run)
```

```
prepare_get_labels
```

Fetch a label for a variable based on its purpose

Description

Fetch a label for a variable based on its purpose

Usage

```
prep_get_labels(
  resp_vars,
  item_level = "item_level",
  label_col,
  max_len = MAX_LABEL_LEN,
  label_class = c("SHORT", "LONG"),
  label_lang = getOption("dataquieR.lang", ""),
  resp_vars_are_var_names_only = FALSE,
  resp_vars_match_label_col_only = FALSE,
  meta_data = item_level,
  meta_data_v2,
  force_label_col = getOption("dataquieR.force_label_col",
    dataquieR.force_label_col_default)
)
```

Arguments

`resp_vars` **variable list** the variable names to fetch for

`item_level` **data.frame** the data frame that contains metadata attributes of study data

`label_col` **variable attribute** the name of the column in the metadata with labels of variables

`max_len` **integer** the maximum label length to return, if not possible w/o causing ambiguous labels, the labels may still be longer

`label_class` **enum** SHORT | LONG. which sort of label according to the metadata model should be returned

`label_lang` **character** optional language suffix, if available in the metadata. Can be controlled by the option `dataquieR.lang`.

`resp_vars_are_var_names_only` **logical** If TRUE, do not use other labels than `VAR_NAMES` for finding `resp_vars` in `meta_data`

`resp_vars_match_label_col_only` **logical** If TRUE, do not use other labels than those, referred by `label_col` for finding `resp_vars` in `meta_data`

`meta_data` **data.frame** old name for `item_level`

`meta_data_v2` **character** path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify `meta_data_v2`.

`force_label_col` **enum** auto | FALSE | TRUE. if TRUE, always use labels according `label_col`, FALSE means use labels matching best the function's requirements, auto means FALSE, if in a `dq_report()` and TRUE, otherwise.

Value

character suitable labels for each `resp_vars`, names of this vector are `VAR_NAMES`

Examples

```
## Not run:
prep_load_workbook_like_file("meta_data_v2")
prep_get_labels("SEX_0", label_class = "SHORT", max_len = 2)

## End(Not run)
```

```
prep_get_study_data_segment
  Get data frame for a given segment
```

Description

Get data frame for a given segment

Usage

```
prep_get_study_data_segment(
  segment,
  study_data,
  item_level = "item_level",
  meta_data = item_level,
  meta_data_v2,
  segment_level,
  meta_data_segment = "segment_level"
)
```

Arguments

segment	character name of the segment to return data for
study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
meta_data	data.frame old name for item_level
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
segment_level	data.frame alias for meta_data_segment
meta_data_segment	data.frame – optional: Segment level metadata

Value

[data.frame](#) the data for the segment

prep_get_user_name *Return the logged-in User's Full Name*

Description

If whoami is not installed, the user name from Sys.info() is returned.

Usage

```
prep_get_user_name()
```

Details

Can be overridden by options or environment:

```
options(FULLNAME = "Stephan Struckmann")
```

```
Sys.setenv(FULLNAME = "Stephan Struckmann")
```

Value

[character](#) the user's name

prep_guess_encoding *Guess encoding of text or text files*

Description

Guess encoding of text or text files

Usage

```
prep_guess_encoding(x, file)
```

Arguments

x [character](#) string to guess encoding for

file [character](#) file to guess encoding for

Value

encoding

prep_link_escape	<i>Prepare a label as part of a link for RMD files</i>
------------------	--

Description

Prepare a label as part of a link for RMD files

Usage

```
prep_link_escape(s, html = FALSE)
```

Arguments

s	the label
html	prepare the label for direct HTML output instead of RMD

Value

the escaped label

prep_list_dataframes	<i>List Loaded Data Frames</i>
----------------------	--------------------------------

Description

List Loaded Data Frames

Usage

```
prep_list_dataframes()
```

Value

names of all loaded data frames

See Also

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_get_data_frame\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_purge_data_frame_cache\(\)](#), [prep_remove_from_cache\(\)](#)

prep_list_voc	<i>All valid voc: vocabularies</i>
---------------	------------------------------------

Description

All valid voc: vocabularies

Usage

```
prep_list_voc()
```

Value

`character()` all voc: suffixes allowed for `prep_get_data_frame()`.

Examples

```
## Not run:
prep_list_dataframes()
prep_list_voc()
prep_get_data_frame("<ICD10>")
my_voc <-
  tibble::tribble(
    ~ voc, ~ url,
    "test", "data:datasets|iris|Species+Sepal.Length")
prep_add_data_frames(`<>` = my_voc)
prep_list_dataframes()
prep_list_voc()
prep_get_data_frame("<test>")
prep_get_data_frame("<ICD10>")
my_voc <-
  tibble::tribble(
    ~ voc, ~ url,
    "ICD10", "data:datasets|iris|Species+Sepal.Length")
prep_add_data_frames(`<>` = my_voc)
prep_list_dataframes()
prep_list_voc()
prep_get_data_frame("<ICD10>")

## End(Not run)
```

`prep_load_folder_with_metadata`*Pre-load a folder with named (usually more than) one table(s)*

Description

These can thereafter be referred to by their names only. Such files are, e.g., spreadsheet-workbooks or RData-files.

Usage

```
prep_load_folder_with_metadata(folder, keep_types = FALSE, ...)
```

Arguments

<code>folder</code>	the folder name to load.
<code>keep_types</code>	logical keep types as possibly defined in the file. set TRUE for study data.
<code>...</code>	arguments passed to []

Details

Note, that this function in contrast to [prep_get_data_frame](#) does neither support selecting specific sheets/columns from a file.

Value

`invisible(the cache environment)`

See Also

[prep_add_data_frames](#)

[prep_get_data_frame](#)

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_get_data_frame\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_purge_data_frame_cache\(\)](#), [prep_remove_from_cache\(\)](#)

`prep_load_report`*Load a dq_report2*

Description

Load a dq_report2

Usage

```
prep_load_report(file)
```

Arguments

file [character](#) the file name to load from

Value

[dataquieR_resultset2](#) the report

```
prep_load_report_from_backend
```

Load a report from a back-end

Description

Load a report from a back-end

Usage

```
prep_load_report_from_backend(
  namespace = "objects",
  db_dir,
  storr_factory = prep_create_storr_factory(namespace = namespace, db_dir = db_dir)
)
```

Arguments

namespace the namespace to read the report's results from

db_dir [character](#) path to the directory for the back-end, if a storr_rds or storr_torr is used.

storr_factory a function returning a storr object holding the report

Value

[dataquieR_resultset2](#) the report

See Also

[prep_create_storr_factory\(\)](#)

Examples

```
## Not run:
r <- dataquieR::dq_report2("study_data", meta_data_v2 = "meta_data_v2",
                           dimensions = NULL)
storr_factory <- prep_create_storr_factory()
r_storr <- prep_set_backend(r, storr_factory)
r_restorr <- prep_set_backend(r_storr, NULL)
r_loaded <- prep_load_report_from_backend(storr_factory)

## End(Not run)
```

`prep_load_workbook_like_file`*Pre-load a file with named (usually more than) one table(s)*

Description

These can thereafter be referred to by their names only. Such files are, e.g., spreadsheet-workbooks or RData-files.

Usage

```
prep_load_workbook_like_file(file, keep_types = FALSE)
```

Arguments

`file` the file name to load.

`keep_types` **logical** keep types as possibly defined in the file. set TRUE for study data.

Details

Note, that this function in contrast to [prep_get_data_frame](#) does neither support selecting specific sheets/columns from a file.

Value

`invisible(the cache environment)`

See Also

[prep_add_data_frames](#)

[prep_get_data_frame](#)

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_get_data_frame\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_purge_data_frame_cache\(\)](#), [prep_remove_from_cache\(\)](#)

`prep_map_labels`*Support function to allocate labels to variables*

Description

Map variables to certain attributes, e.g. by default their labels.

Usage

```
prep_map_labels(
  x,
  item_level = "item_level",
  to = LABEL,
  from = VAR_NAMES,
  ifnotfound,
  warn_ambiguous = FALSE,
  meta_data_v2,
  meta_data = item_level
)
```

Arguments

x	character variable names, character vector, see parameter from
item_level	data.frame metadata data frame, if, as a dataquieR developer, you do not have item-level-metadata , you should use util_map_labels instead to avoid consistency checks on for item-level meta_data.
to	character variable attribute to map to
from	character variable identifier to map from
ifnotfound	list A list of values to be used if the item is not found: it will be coerced to a list if necessary.
warn_ambiguous	logical print a warning if mapping variables from from to to produces ambiguous identifiers.
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify meta_data_v2.
meta_data	data.frame old name for item_level

Details

This function basically calls `colnames(study_data) <- meta_data$LABEL`, ensuring correct merging/joining of study data columns to the corresponding metadata rows, even if the orders differ. If a variable/study_data-column name is not found in `meta_data[[from]]` (default from = VAR_NAMES), either stop is called or, if ifnotfound has been assigned a value, that value is returned. See [mget](#), which is internally used by this function.

The function not only maps to the LABEL column, but to can be any metadata variable attribute, so the function can also be used, to get, e.g. all HARD_LIMITS from the metadata.

Value

a character vector with:

- mapped values

Examples

```
## Not run:
meta_data <- prep_create_meta(
  VAR_NAMES = c("ID", "SEX", "AGE", "DOE"),
  LABEL = c("Pseudo-ID", "Gender", "Age", "Examination Date"),
  DATA_TYPE = c(DATA_TYPES$INTEGER, DATA_TYPES$INTEGER, DATA_TYPES$INTEGER,
                 DATA_TYPES$DATETIME),
  MISSING_LIST = ""
)
stopifnot(all(prepare_map_labels(c("AGE", "DOE"), meta_data) == c("Age",
                                                                "Examination Date")))

## End(Not run)
```

prep_merge_study_data *Merge a list of study data frames to one (sparse) study data frame*

Description

Merge a list of study data frames to one (sparse) study data frame

Usage

```
prep_merge_study_data(study_data_list)
```

Arguments

```
study_data_list
  list the list
```

Value

```
data.frame study_data
```

prep_meta_data_v1_to_item_level_meta_data
Convert item-level metadata from v1.0 to v2.0

Description

This function is idempotent..

Usage

```
prep_meta_data_v1_to_item_level_meta_data(
  item_level = "item_level",
  verbose = TRUE,
  label_col = LABEL,
  cause_label_df,
  meta_data = item_level
)
```

Arguments

`item_level` [data.frame](#) the old item-level-metadata

`verbose` [logical](#) display all estimated decisions, defaults to TRUE, except if called in a `dq_report2` pipeline.

`label_col` [variable attribute](#) the name of the column in the metadata with labels of variables

`cause_label_df` [data.frame](#) missing code table, see [cause_label_df](#). Optional. If this argument is given, you can add missing code tables.

`meta_data` [data.frame](#) old name for `item_level`

Details

The `options("dataquieR.force_item_specific_missing_codes")` (default FALSE) tells the system, to always fill in `res_vars` columns to the `MISSING_LIST_TABLE`, even, if the column already exists, but is empty.

Value

[data.frame](#) the updated metadata

<code>prep_min_obs_level</code>	<i>Support function to identify the levels of a process variable with minimum number of observations</i>
---------------------------------	--

Description

utility function to subset data based on minimum number of observation per level

Usage

```
prep_min_obs_level(study_data, group_vars, min_obs_in_subgroup)
```


Arguments

- study_data [data.frame](#) the data frame that contains the measurements
- group_vars [variable list](#) the name grouping variable
- min_obs_in_subgroup
[integer](#) optional argument if a "group_var" is used. This argument specifies the minimum no. of observations that is required to include a subgroup (level) of the "group_var" in the analysis. Subgroups with less observations are excluded. The default is 30.

Details

This functions removes observations having fewer than min_obs_in_subgroup distinct values in a group variable, e.g. blood pressure measurements performed by an examiner having fewer than e.g. 50 measurements done. It displays a warning, if samples/rows are removed and returns the modified study data frame.

Value

a data frame with:

- a subsample of original data

prep_open_in_excel *Open a data frame in Excel*

Description

Open a data frame in Excel

Usage

```
prep_open_in_excel(dfr)
```

Arguments

dfr the data frame

Details

if the file cannot be read on function exit, NULL will be returned

Value

potentially modified data frame after dialog was closed

```
prep_pmap
```

Support function for a parallel pmap

Description

parallel version of `purrr::pmap`

Usage

```
prep_pmap(.l, .f, ..., cores = 0)
```

Arguments

`.l` [data.frame](#) with one call per line and one function argument per column

`.f` [function](#) to call with the arguments from `.l`

`...` additional, static arguments for calling `.f`

`cores` number of cpu cores to use or a (named) list with arguments for [parallelMap::parallelStart](#) or NULL, if parallel has already been started by the caller. Set to 0 to run without parallelization.

Value

[list](#) of results of the function calls

Author(s)

[Aurèle](#)
S Struckmann

See Also

`purrr::pmap`
[Stack Overflow post](#)

```
prep_prepare_dataframes
```

Prepare and verify study data with metadata

Description

This function ensures, that a data frame `ds1` with suitable variable names `study_data` and `meta_data` exist as base [data.frames](#).

Usage

```
prep_prepare_dataframes(
  .study_data,
  .meta_data,
  .label_col,
  .replace_hard_limits,
  .replace_missings,
  .sm_code = NULL,
  .allow_empty = FALSE,
  .adjust_data_type = TRUE,
  .amend_scale_level = TRUE,
  .apply_factor_metadata = FALSE,
  .apply_factor_metadata_inadm = FALSE,
  .internal = rlang::env_inherits(rlang::caller_env(), parent.env(environment()))
)
```

Arguments

<code>.study_data</code>	if provided, use this data set as <code>study_data</code>
<code>.meta_data</code>	if provided, use this data set as <code>meta_data</code>
<code>.label_col</code>	if provided, use this as <code>label_col</code>
<code>.replace_hard_limits</code>	replace HARD_LIMIT violations by NA, defaults to FALSE.
<code>.replace_missings</code>	replace missing codes, defaults to TRUE
<code>.sm_code</code>	missing code for NAs, if they have been re-coded by <code>util_combine_missing_lists</code>
<code>.allow_empty</code>	allow ds1 to be empty, i.e., 0 rows and/or 0 columns
<code>.adjust_data_type</code>	ensure that the data type of variables in the study data corresponds to their data type specified in the metadata
<code>.amend_scale_level</code>	ensure that SCALE_LEVEL is available in the item-level <code>meta_data</code> . internally used to prevent recursion, if called from <code>prep_scalelevel_from_data_and_metadata()</code> .
<code>.apply_factor_metadata</code>	logical convert categorical variables to labeled factors.
<code>.apply_factor_metadata_inadm</code>	logical convert categorical variables to labeled factors keeping inadmissible values. Implies, that <code>.apply_factor_metadata</code> will be set to TRUE, too.
<code>.internal</code>	logical internally called, modify caller's environment.

Details

This function defines `ds1` and modifies `study_data` and `meta_data` in the environment of its caller (see [eval.parent\(\)](#)). It also defines or modifies the object `label_col` in the calling environment. Almost all functions exported by `dataquieR` call this function initially, so that aspects common to all functions live here, e.g. testing, if an argument `meta_data` has been given and features really a

`data.frame`. It verifies the existence of required metadata attributes (`VARATT_REQUIRE_LEVELS`). It can also replace missing codes by NAs, and calls `prep_study2meta` to generate a minimum set of metadata from the study data on the fly (should be amended, so on-the-fly-calling is not recommended for an instructive use of `dataquieR`).

The function also detects tibbles, which are then converted to base-R `data.frames`, which are expected by `dataquieR`.

If `.internal` is `TRUE`, differently from the other utility function that work in their caller's environment, this function modifies objects in the calling function's environment. It defines a new object `ds1`, it modifies `study_data` and/or `meta_data` and `label_col`.

Value

`ds1` the study data with mapped column names

See Also

`acc_margins`

Examples

```
## Not run:
acc_test1 <- function(resp_variable, aux_variable,
                      time_variable, co_variables,
                      group_vars, study_data, meta_data) {
  prep_prepare_dataframes()
  invisible(ds1)
}
acc_test2 <- function(resp_variable, aux_variable,
                      time_variable, co_variables,
                      group_vars, study_data, meta_data, label_col) {
  ds1 <- prep_prepare_dataframes(study_data, meta_data)
  invisible(ds1)
}
environment(acc_test1) <- asNamespace("dataquieR")
# perform this inside the package (not needed for functions that have been
# integrated with the package already)

environment(acc_test2) <- asNamespace("dataquieR")
# perform this inside the package (not needed for functions that have been
# integrated with the package already)
acc_test3 <- function(resp_variable, aux_variable, time_variable,
                      co_variables, group_vars, study_data, meta_data,
                      label_col) {
  prep_prepare_dataframes()
  invisible(ds1)
}
acc_test4 <- function(resp_variable, aux_variable, time_variable,
                      co_variables, group_vars, study_data, meta_data,
                      label_col) {
  ds1 <- prep_prepare_dataframes(study_data, meta_data)
  invisible(ds1)
}
```

```

}
environment(acc_test3) <- asNamespace("dataquieR")
# perform this inside the package (not needed for functions that have been
# integrated with the package already)

environment(acc_test4) <- asNamespace("dataquieR")
# perform this inside the package (not needed for functions that have been
# integrated with the package already)
meta_data <- prep_get_data_frame("meta_data")
study_data <- prep_get_data_frame("study_data")
try(acc_test1())
try(acc_test2())
acc_test1(study_data = study_data)
try(acc_test1(meta_data = meta_data))
try(acc_test2(study_data = 12, meta_data = meta_data))
print(head(acc_test1(study_data = study_data, meta_data = meta_data)))
print(head(acc_test2(study_data = study_data, meta_data = meta_data)))
print(head(acc_test3(study_data = study_data, meta_data = meta_data)))
print(head(acc_test3(study_data = study_data, meta_data = meta_data,
  label_col = LABEL)))
print(head(acc_test4(study_data = study_data, meta_data = meta_data)))
print(head(acc_test4(study_data = study_data, meta_data = meta_data,
  label_col = LABEL)))
try(acc_test2(study_data = NULL, meta_data = meta_data))

## End(Not run)

```

```

prep_purge_data_frame_cache
      Clear data frame cache

```

Description

Clear data frame cache

Usage

```
prep_purge_data_frame_cache()
```

Value

nothing

See Also

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_get_data_frame\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_remove_from_cache\(\)](#)

`prep_remove_from_cache`*Remove a specified element from the data frame cache*

Description

Remove a specified element from the data frame cache

Usage

```
prep_remove_from_cache(object_to_remove)
```

Arguments

`object_to_remove`

character name of the object to be removed as character string (quoted), or character vector containing the names of the objects to remove from the cache

Value

nothing

See Also

Other data-frame-cache: [prep_add_data_frames\(\)](#), [prep_get_data_frame\(\)](#), [prep_list_dataframes\(\)](#), [prep_load_folder_with_metadata\(\)](#), [prep_load_workbook_like_file\(\)](#), [prep_purge_data_frame_cache\(\)](#)

Examples

```
## Not run:
prep_load_workbook_like_file("meta_data_v2") #load metadata in the cache
ls(.dataframe_environment()) #get the list of dataframes in the cache

#remove cross-item_level from the cache
prep_remove_from_cache("cross-item_level")

#remove dataframe_level and expected_id from the cache
prep_remove_from_cache(c("dataframe_level", "expected_id"))

#remove missing_table and segment_level from the cache
x<- c("missing_table", "segment_level")
prep_remove_from_cache(x)

## End(Not run)
```

prep_render_pie_chart_from_summaryclasses_ggplot2
Create a ggplot2 pie chart

Description

Create a ggplot2 pie chart

Usage

```
prep_render_pie_chart_from_summaryclasses_ggplot2(  
  data,  
  meta_data = "item_level"  
)
```

Arguments

data	data as returned by prep_summary_to_classes but summarized by one column (currently, we support indicator_metric, STUDY_SEGMENT, and VAR_NAMES)
meta_data	meta_data

Value

a [ggplot2::ggplot2](#) plot

See Also

Other summary_functions: [prep_combine_report_summaries\(\)](#), [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_extract_summary.dataquieR_result.dataquieR_result\(\)](#), [prep_render_pie_chart_from_summaryclasses_plotly\(\)](#), [prep_summary_to_classes\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

prep_render_pie_chart_from_summaryclasses_plotly
Create a plotly pie chart

Description

Create a plotly pie chart

Usage

```
prep_render_pie_chart_from_summaryclasses_plotly(
  data,
  meta_data = "item_level"
)
```

Arguments

data	data as returned by <code>prep_summary_to_classes</code> but summarized by one column (currently, we support <code>indicator_metric</code> , <code>call_names</code> , <code>STUDY_SEGMENT</code> , and <code>VAR_NAMES</code>)
meta_data	meta_data

Value

a `htmltools` compatible object

See Also

Other `summary_functions`: [prep_combine_report_summaries\(\)](#), [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_extract_summary.dataquieR_result.dataquieR_result\(\)](#), [prep_render_pie_chart_from_summaryclasses_ggplot2\(\)](#), [prep_summary_to_classes\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

`prep_robust_guess_data_type`

Guess the data type of a vector

Description

Guess the data type of a vector

Usage

```
prep_robust_guess_data_type(x, k = 50, it = 200)
```

Arguments

x	a vector with characters
k	numeric sample size, if less than <code>floor(length(x) / (it/20))</code> , minimum sample size is 1.
it	integer number of iterations when taking samples

Value

a guess of the data type of `x`. An attribute `orig_type` is also attached to give the more detailed guess returned by `readr::guess_parser()`.

Algorithm

This function takes `x` and tries to guess the data type of random subsets of this vector using `readr::guess_parser()`. The RNG is initialized with a constant, so the function stays deterministic. It does such sub-sample based checks it times, the majority of the detected datatype determines the guessed data type.

prep_save_report	Save a dq_report2
------------------	-------------------

Description

Save a dq_report2

Usage

```
prep_save_report(report, file, compression_level = 3)
```

Arguments

`report` [dataquieR_resultset2](#) the report
`file` [character](#) the file name to write to
`compression_level` [integer](#) from=0 to=9. Compression level. 9 is very slow.

Value

`invisible(NULL)`

prep_scalelevel_from_data_and_metadata	<i>Heuristics to amend a SCALE_LEVEL column and a UNIT column in the metadata</i>
--	---

Description

...if missing

Usage

```

prep_scalelevel_from_data_and_metadata(
  resp_vars = lifecycle::deprecated(),
  study_data,
  item_level = "item_level",
  label_col = LABEL,
  meta_data = item_level,
  meta_data_v2
)

```

Arguments

resp_vars [variable list](#) deprecated, the function always addresses all variables.
study_data [data.frame](#) the data frame that contains the measurements
item_level [data.frame](#) the data frame that contains metadata attributes of study data
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
meta_data [data.frame](#) old name for item_level
meta_data_v2 [character](#) path to workbook like metadata file, see [prep_load_workbook_like_file](#) for details. **ALL LOADED DATAFRAMES WILL BE PURGED**, using [prep_purge_data_frame_cache](#), if you specify meta_data_v2.

Value

[data.frame](#) modified metadata

Examples

```

## Not run:
  prep_load_workbook_like_file("meta_data_v2")
  prep_scalelevel_from_data_and_metadata(study_data = "study_data")

## End(Not run)

```

prep_set_backend *Change the back-end of a report*

Description

with this function, you can move a report from/to a storrr storage.

Usage

```

prep_set_backend(r, storrr_factory = NULL, amend = FALSE)

```

Arguments

`r` [dataquieR_resultset2](#) the report

`storr_factory` `storr` the `storr` storage or `NULL`, to move the report fully back into the RAM.

`amend` [logical](#) if there is already data in `storr_factory`, use it anyways – unsupported, so far!

Value

[dataquieR_resultset2](#) but now with the desired back-end

prep_study2meta	<i>Guess a metadata data frame from study data.</i>
-----------------	---

Description

Guess a minimum metadata data frame from study data. Minimum required variable attributes are:

Usage

```
prep_study2meta(
  study_data,
  level = c(VARATT_REQUIRE_LEVELS$REQUIRED, VARATT_REQUIRE_LEVELS$RECOMMENDED),
  cumulative = TRUE,
  convert_factors = FALSE,
  guess_missing_codes = getOption("dataquieR.guess_missing_codes",
    dataquieR.guess_missing_codes_default)
)
```

Arguments

`study_data` [data.frame](#) the data frame that contains the measurements

`level` [enum](#) levels to provide (see also [VARATT_REQUIRE_LEVELS](#))

`cumulative` [logical](#) include attributes of all levels up to level

`convert_factors` [logical](#) convert factor columns to coded integers. if selected, then also the study data will be updated and returned.

`guess_missing_codes` [logical](#) try to guess missing codes from the data

Details

```
dataquieR:::util_get_var_att_names_of_level(VARATT_REQUIRE_LEVELS$REQUIRED)
#>      VAR_NAMES      DATA_TYPE MISSING_LIST_TABLE
#>      "VAR_NAMES"      "DATA_TYPE" "MISSING_LIST_TABLE"
```

The function also tries to detect missing codes.

Value

a meta_data data frame or a list with study data and metadata, if convert_factors == TRUE.

Examples

```
## Not run:  
dataquieR::prep_study2meta(Orange, convert_factors = FALSE)  
  
## End(Not run)
```

prep_summary_to_classes

Classify metrics from a report summary table

Description

Classify metrics from a report summary table

Usage

```
prep_summary_to_classes(report_summary)
```

Arguments

report_summary [list\(\)](#) as returned by [prep_extract_summary\(\)](#)

Value

[data.frame](#) classes for the report summary table, long format

See Also

Other summary_functions: [prep_combine_report_summaries\(\)](#), [prep_extract_classes_by_functions\(\)](#), [prep_extract_summary\(\)](#), [prep_extract_summary.dataquieR_result\(\)](#), [prep_extract_summary.dataquieR_result.prep_render_pie_chart_from_summaryclasses_ggplot2\(\)](#), [prep_render_pie_chart_from_summaryclasses_plot1\(\)](#), [util_as_cat\(\)](#), [util_as_integer_cat\(\)](#), [util_extract_indicator_metrics\(\)](#), [util_get_category_for_result\(\)](#), [util_get_colors\(\)](#), [util_get_labels_grading_class\(\)](#), [util_get_message_for_result\(\)](#), [util_get_rule_sets\(\)](#), [util_get_ruleset_formats\(\)](#), [util_get_thresholds\(\)](#), [util_html_table\(\)](#), [util_sort_by_order\(\)](#)

prep_title_escape *Prepare a label as part of a title text for RMD files*

Description

Prepare a label as part of a title text for RMD files

Usage

```
prep_title_escape(s, html = FALSE)
```

Arguments

s the label
html prepare the label for direct HTML output instead of RMD

Value

the escaped label

prep_undisclose *Remove data disclosing details*

Description

new function: no warranty, so far.

Usage

```
prep_undisclose(x)
```

Arguments

x an object to un-disclose, a

Value

undisclosed object

prep_unsplit_val_tabs *Combine all missing and value lists to one big table*

Description

Combine all missing and value lists to one big table

Usage

```
prep_unsplit_val_tabs(meta_data = "item_level", val_tab = NULL)
```

Arguments

meta_data [data.frame](#) item level meta data to be used, defaults to "item_level"
val_tab [character](#) name of the table being created: This table will be added to the data frame cache (or overwritten). If NULL, the table will only be returned

Value

[data.frame](#) the combined table

prep_valuelabels_from_data
Get value labels from data

Description

Detects factors and converts them to compatible metadata/study data.

Usage

```
prep_valuelabels_from_data(resp_vars = colnames(study_data), study_data)
```

Arguments

resp_vars [variable](#) names of the variables to fetch the value labels from the data
study_data [data.frame](#) the data frame that contains the measurements

Value

a [list](#) with:

- VALUE_LABELS: vector of value labels and modified study data
- ModifiedStudyData: study data with factors as integers

Examples

```
## Not run:  
dataquieR::prep_datatype_from_data(iris)  
  
## End(Not run)
```

```
print.dataquieR_result
```

Print a [dataquieR](#) result returned by [dq_report2](#)

Description

Print a [dataquieR](#) result returned by [dq_report2](#)

Usage

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

Arguments

x	list a dataquieR result from dq_report2 or util_eval_to_dataquieR_result
...	passed to print. Additionally, the argument slot may be passed to print only specific sub-results.

Value

see print

See Also

[util_pretty_print\(\)](#)

```
print.dataquieR_resultset
```

Generate a RMarkdown-based report from a [dataquieR](#) report

Description

Generate a RMarkdown-based report from a [dataquieR](#) report

Usage

```
## S3 method for class 'dataquieR_resultset'  
print(...)
```

Arguments

... deprecated

Value

deprecated

```
print.dataquieR_resultset2
```

Generate a HTML-based report from a [dataquieR](#) report

Description

Generate a HTML-based report from a [dataquieR](#) report

Usage

```
## S3 method for class 'dataquieR_resultset2'
print(
  x,
  dir,
  view = TRUE,
  disable_plotly = FALSE,
  block_load_factor = 4,
  advanced_options = list(),
  dashboard = NA,
  ...
)
```

Arguments

x [dataquieR](#) report v2.

dir [character](#) directory to store the rendered report's files, a temporary one, if omitted. Directory will be created, if missing, files may be overwritten inside that directory

view [logical](#) display the report

disable_plotly [logical](#) do not use plotly, even if installed

block_load_factor [numeric](#) multiply size of parallel compute blocks by this factor.

advanced_options [list](#) options to set during report computation, see [options\(\)](#)

dashboard [logical](#) dashboard mode: TRUE: create a dashboard only, FALSE: don't create a dashboard at all, NA or missing: create a "normal" report with a dashboard included.

... additional arguments:

Value

file names of the generated report's HTML files

```
print.dataquieR_summary
```

Print a dataquieR summary

Description

Print a dataquieR summary

Usage

```
## S3 method for class 'dataquieR_summary'
print(
  x,
  ...,
  grouped_by = c("call_names", "indicator_metric"),
  dont_print = FALSE,
  folder_of_report = NULL
)
```

Arguments

x	the dataquieR summary, see summary() and dq_report2()
...	not yet used
grouped_by	define the columns of the resulting matrix. It can be either "call_names", one column per function, or "indicator_metric", one column per indicator or both c("call_names", "indicator_metric"). The last combination is the default
dont_print	suppress the actual printing, just return a printable object derived from x
folder_of_report	a named vector with the location of variable and call_names

Value

invisible html object

`print.DataSlot` *Print a DataSlot object*

Description

Print a DataSlot object

Usage

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

Arguments

<code>x</code>	the object
<code>...</code>	not used

Value

see `print`

`print.interval` *print implementation for the class interval*

Description

such objects, for now, only occur in RECCap rules, so this function is meant for internal use, mostly – for now.

Usage

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

Arguments

<code>x</code>	interval objects to print
<code>...</code>	not used yet

Value

the printed object

See Also

`base::print`

print.list	<i>print a list of dataquieR_result objects</i>
------------	---

Description

print a list of dataquieR_result objects

Usage

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

Arguments

x	list() only, if all elements inherit from dataquieR_result , this implementation runs
...	passed to other implementations

Value

undefined

print.master_result	<i>Print a master_result object</i>
---------------------	-------------------------------------

Description

Print a master_result object

Usage

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

Arguments

x	the object
...	not used

Value

invisible(NULL)

```
print.ReportSummaryTable
    print implementation for the class ReportSummaryTable
```

Description

Use this function to print results objects of the class ReportSummaryTable.

Usage

```
## S3 method for class 'ReportSummaryTable'
print(
  x,
  relative = lifecycle::deprecated(),
  dt = FALSE,
  fillContainer = FALSE,
  displayValues = FALSE,
  view = TRUE,
  ...,
  flip_mode = "auto"
)
```

Arguments

x	ReportSummaryTable objects to print
relative	deprecated
dt	logical use DT::datatables, if installed
fillContainer	logical if dt is TRUE, control table size, see DT::datatables.
displayValues	logical if dt is TRUE, also display the actual values
view	logical if view is FALSE, do not print but return the output, only
...	not used, yet
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the roptions(dataquieR.flip_mode = ...). If called from dq_report, you can also pass flip_mode to all function calls or set them specifically using specific_args.

Value

the printed object

See Also

base::print

print.Slot	<i>Print a Slot object</i>
------------	----------------------------

Description

displays all warnings and stuff. then it prints x.

Usage

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

Arguments

x	the object
...	not used

Value

calls the next print method

print.StudyDataSlot	<i>Print a StudyDataSlot object</i>
---------------------	-------------------------------------

Description

Print a StudyDataSlot object

Usage

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

Arguments

x	the object
...	not used

Value

see print

print.TableSlot *Print a TableSlot object*

Description

Print a TableSlot object

Usage

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

Arguments

x	the object
...	not used

Value

see print

pro_applicability_matrix
Check applicability of DQ functions on study data

Description

Checks applicability of DQ functions based on study data and metadata characteristics

Usage

```
pro_applicability_matrix(  
  study_data,  
  item_level = "item_level",  
  split_segments = FALSE,  
  label_col,  
  max_vars_per_plot = 20,  
  meta_data_segment,  
  meta_data_dataframe,  
  flip_mode = "noflip",  
  meta_data_v2,  
  meta_data = item_level,  
  segment_level,  
  dataframe_level  
)
```

Arguments

study_data	data.frame the data frame that contains the measurements
item_level	data.frame the data frame that contains metadata attributes of study data
split_segments	logical return one matrix per study segment
label_col	variable attribute the name of the column in the metadata with labels of variables
max_vars_per_plot	integer from=0. The maximum number of variables per single plot.
meta_data_segment	data.frame – optional: Segment level metadata
meta_data_dataframe	data.frame – optional: Data frame level metadata
flip_mode	enum default flip noflip auto. Should the plot be in default orientation, flipped, not flipped or auto-flipped. Not all options are always supported. In general, this can be controlled by setting the <code>roptions(dataquieR.flip_mode = ...)</code> . If called from <code>dq_report</code> , you can also pass <code>flip_mode</code> to all function calls or set them specifically using <code>specific_args</code> .
meta_data_v2	character path to workbook like metadata file, see prep_load_workbook_like_file for details. ALL LOADED DATAFRAMES WILL BE PURGED , using prep_purge_data_frame_cache , if you specify <code>meta_data_v2</code> .
meta_data	data.frame old name for <code>item_level</code>
segment_level_dataframe_level	data.frame alias for <code>meta_data_segment</code>
	data.frame alias for <code>meta_data_dataframe</code>

Details

This is a preparatory support function that compares study data with associated metadata. A prerequisite of this function is that the no. of columns in the study data complies with the no. of rows in the metadata.

For each existing R-implementation, the function searches for necessary static metadata and returns a heatmap like matrix indicating the applicability of each data quality implementation.

In addition, the data type defined in the metadata is compared with the observed data type in the study data.

Value

a list with:

- `SummaryTable`: data frame about the applicability of each indicator function (each function in a column). its [integer](#) values can be one of the following four categories: 0. Non-matching datatype + Incomplete metadata, 1. Non-matching datatype + complete metadata, 2. Matching datatype + Incomplete metadata, 3. Matching datatype + complete metadata, 4. Not applicable according to data type
- `ApplicabilityPlot`: [ggplot2::ggplot2](#) heatmap plot, graphical representation of `SummaryTable`
- `ApplicabilityPlotList`: [list](#) of plots per (maybe artificial) segment
- `ReportSummaryTable`: data frame underlying `ApplicabilityPlot`

`rbind.ReportSummaryTable`
Combine ReportSummaryTable outputs

Description

Using this `rbind` implementation, you can combine different heatmap-like results of the class `ReportSummaryTable`.

Usage

```
## S3 method for class 'ReportSummaryTable'
rbind(...)
```

Arguments

... ReportSummaryTable objects to combine.

See Also

[base::rbind.data.frame](#)

`REL_VAL` *Cross-item level metadata attribute name*

Description

Specifies the type of reliability or validity analysis. The string specifies the analysis algorithm to be used, and can be either "inter-class" or "intra-class".

Usage

```
REL_VAL
```

Format

An object of class character of length 1.

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [VARIABLE_LIST](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

resnames	<i>Return names of result slots (e.g., 3rd dimension of dataquieR results)</i>
----------	--

Description

Return names of result slots (e.g., 3rd dimension of dataquieR results)

Usage

```
resnames(x)
```

Arguments

x the objects

Value

character vector with names

resnames.dataquieR_resultset2	<i>Return names of result slots (e.g., 3rd dimension of dataquieR results)</i>
-------------------------------	--

Description

Return names of result slots (e.g., 3rd dimension of dataquieR results)

Usage

```
## S3 method for class 'dataquieR_resultset2'  
resnames(x)
```

Arguments

x the objects

Value

character vector with names

SCALE_LEVELS	<i>Scale Levels</i>
--------------	---------------------

Description

Scale Levels of Study Data according to Stevens' s Typology:

In the metadata, the following entries are allowed for the [variable attribute SCALE_LEVEL](#):

Usage

SCALE_LEVELS

Format

An object of class `list` of length 5.

Details

- `nominal` for categorical variables
- `ordinal` for ordinal variables (i.e., comparison of values is possible)
- `interval` for interval scales, i.e., distances are meaningful
- `ratio` for ratio scales, i.e., ratios are meaningful
- `na` for variables, that contain e.g. unstructured texts, `json`, `xml`, ... to distinguish them from variables, that still need to have the `SCALE_LEVEL` estimated by `prep_scalelevel_from_data_and_metadata()`

Examples:

- sex, eye color – nominal
- income group, education level – ordinal
- temperature in degree Celsius – interval
- body weight, temperature in Kelvin – ratio

See Also

[Wikipedia](#)

SEGMENT_ID_REF_TABLE *Segment level metadata attribute name*

Description

The name of the data frame containing the reference IDs to be compared with the IDs in the targeted segment.

Usage

SEGMENT_ID_REF_TABLE

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_ID_TABLE *Deprecated segment level metadata attribute name*

Description

The name of the data frame containing the reference IDs to be compared with the IDs in the targeted segment.

Usage

SEGMENT_ID_TABLE

Format

An object of class character of length 1.

Details

Please use [SEGMENT_ID_REF_TABLE](#)

SEGMENT_ID_VARS	<i>Segment level metadata attribute name</i>
-----------------	--

Description

All variables that are to be used as one single ID variable (combined key) in a segment.

Usage

SEGMENT_ID_VARS

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_MISS	<i>Segment level metadata attribute name</i>
--------------	--

Description

true or false to suppress crude segment missingness output (Completeness/Misg. Segments in the report). Defaults to compute the output, if more than one segment is available in the item-level metadata.

Usage

SEGMENT_MISS

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_PART_VARS	<i>Segment level metadata attribute name</i>
-------------------	--

Description

The name of the segment participation status variable

Usage

SEGMENT_PART_VARS

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_RECORD_CHECK	<i>Segment level metadata attribute name</i>
----------------------	--

Description

The type of check to be conducted when comparing the reference ID table with the IDs in a segment.

Usage

SEGMENT_RECORD_CHECK

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_RECORD_COUNT *Segment level metadata attribute name*

Description

Number of expected data records in each segment. [numeric](#). Check only conducted if number entered

Usage

SEGMENT_RECORD_COUNT

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SEGMENT_UNIQUE_ID *Segment level metadata attribute name*

Description

Segment level metadata attribute name

Usage

SEGMENT_UNIQUE_ID

Format

An object of class character of length 1.

See Also

[DF_UNIQUE_ID](#)
[meta_data_segment](#)

SEGMENT_UNIQUE_ROWS	<i>Segment level metadata attribute name</i>
---------------------	--

Description

Specifies whether identical data is permitted across rows in a segment (excluding ID variables)

Usage

SEGMENT_UNIQUE_ROWS

Format

An object of class character of length 1.

See Also

[meta_data_segment](#)

SPLIT_CHAR	<i>Character used by default as a separator in metadata such as missing codes</i>
------------	---

Description

This 1 character is according to our metadata concept "I".

Usage

SPLIT_CHAR

Format

An object of class character of length 1.

study_data	<i>Data frame with the study data whose quality is being assessed</i>
------------	---

Description

Study data is expected in wide format. It should contain all variables for all segments in one large table, even, if some variables are not measured for all observational units (study participants).

```
summary.dataquieR_resultset
      Summarize a dataquieR report
```

Description

Deprecated

Usage

```
## S3 method for class 'dataquieR_resultset'
summary(...)
```

Arguments

... Deprecated

Value

Deprecated

```
summary.dataquieR_resultset2
      Generate a report summary table
```

Description

Generate a report summary table

Usage

```
## S3 method for class 'dataquieR_resultset2'
summary(
  object,
  aspect = c("applicability", "error", "anamat", "indicator_or_descriptor"),
  FUN,
  collapse = "\n<br />\n",
  ...
)
```

Arguments

object	a square result set
aspect	an aspect/problem category of results
FUN	function to apply to the cells of the result table
collapse	passed to FUN
...	not used

Value

a summary of a dataquieR report

Examples

```
## Not run:
  util_html_table(summary(report),
    filter = "top", options = list(scrollCollapse = TRUE, scrolly = "75vh"),
    is_matrix_table = TRUE, rotate_headers = TRUE, output_format = "HTML"
  )

## End(Not run)
```

UNITS

Valid unit symbols according to `units::valid_udunits()`

Description

like m, g, N, ...

See Also

Other UNITS: [UNIT_IS_COUNT](#), [UNIT_PREFIXES](#), [UNIT_SOURCES](#), [WELL_KNOWN_META_VARIABLE_NAMES](#)

UNIT_IS_COUNT

Is a unit a count according to `units::valid_udunits()`

Description

see column def, therein

Details

like %, ppt, ppm

See Also

Other UNITS: [UNITS](#), [UNIT_PREFIXES](#), [UNIT_SOURCES](#), [WELL_KNOWN_META_VARIABLE_NAMES](#)

UNIT_PREFIXES	<i>Valid unit prefixes according to</i> <code>units::valid_udunits_prefixes()</code>
---------------	---

Description

like k, m, M, c, ...

See Also

Other UNITS: [UNITS](#), [UNIT_IS_COUNT](#), [UNIT_SOURCES](#), [WELL_KNOWN_META_VARIABLE_NAMES](#)

UNIT_SOURCES	<i>Maturity stage of a unit according to</i> <code>units::valid_udunits()</code>
--------------	--

Description

see column `source_xml` therein, i.e., base, derived, accepted, or common

See Also

Other UNITS: [UNITS](#), [UNIT_IS_COUNT](#), [UNIT_PREFIXES](#), [WELL_KNOWN_META_VARIABLE_NAMES](#)

UNIVARIATE_OUTLIER_CHECKTYPE	<i>Item level metadata attribute name</i>
------------------------------	---

Description

Select, which outlier criteria to compute, see [acc_univariate_outlier](#).

Usage

UNIVARIATE_OUTLIER_CHECKTYPE

Format

An object of class character of length 1.

Details

You can leave the cell empty, then, all checks will apply. If you enter a set of methods, the maximum for [N_RULES](#) changes. See also [MULTIVARIATE_OUTLIER_CHECKTYPE](#).

See Also

[WELL_KNOWN_META_VARIABLE_NAMES](#)

`util_bar_plot`*Utility function to create bar plots*

Description

A helper function for simple bar plots. The layout is intended for data with positive numbers only (e.g., counts/frequencies).

Usage

```
util_bar_plot(  
  plot_data,  
  cat_var,  
  num_var,  
  relative = FALSE,  
  show_numbers = TRUE,  
  fill_var = NULL,  
  colors = "#2166AC",  
  show_color_legend = FALSE,  
  flip = FALSE  
)
```

Arguments

<code>plot_data</code>	the data for the plot. It should consist of one column specifying the categories, and a second column giving the respective numbers / counts per category. It may contain another column to specify the coloring of the bars (<code>fill_var</code>).
<code>cat_var</code>	column name of the categorical variable in <code>plot_data</code>
<code>num_var</code>	column name of the numerical variable in <code>plot_data</code>
<code>relative</code>	if TRUE, numbers will be interpreted as percentages (values in <code>num_var</code> should lie within $[0, 1]$)
<code>show_numbers</code>	if TRUE, numbers will be displayed on top of the bars
<code>fill_var</code>	column name of the variable in <code>plot_data</code> which will be used to color the bars individually
<code>colors</code>	vector of colors, or a single color
<code>show_color_legend</code>	if TRUE, a legend for the colors will be displayed
<code>flip</code>	if TRUE, bars will be oriented horizontally

Value

a bar plot

util_combine_list_report_summaries

Create a data frame containing all the results from summaries of reports

Description

Create a data frame containing all the results from summaries of reports

Usage

```
util_combine_list_report_summaries(  
  to_combine,  
  type = c("unique_vars", "repeated_vars")  
)
```

Arguments

`to_combine` **vector** a list containing the summaries of reports obtained with `summary(report)`

`type` **character** if type is `unique_vars` it means that the variable names are unique and there is not need to add a prefix to the variables and labels to specify the report of origin. If type is `repeated_vars` a prefix will be used to specify the report of origin of each variable

Value

a summary of summaries of dataquieR reports

util_compute_kurtosis *Compute Kurtosis*

Description

Compute Kurtosis

Usage

```
util_compute_kurtosis(x)
```

Arguments

`x` **data**

Value

the Kurtosis

util_compute_SE_skewness
Compute SE.Skewness

Description

Compute SE.Skewness

Usage

```
util_compute_SE_skewness(x, skewness = util_compute_skewness(x))
```

Arguments

x	data
skewness	if already known

Value

the standard error of skewness

util_compute_skewness *Compute the Skewness*

Description

Compute the Skewness

Usage

```
util_compute_skewness(x)
```

Arguments

x	data
---	------

Value

the Skewness

```
util_create_report_by_overview
```

Create an overview of the reports created with dq_report_by

Description

Create an overview of the reports created with dq_report_by

Usage

```
util_create_report_by_overview(  
  output_dir,  
  strata_column,  
  segment_column,  
  strata_column_label,  
  subgroup,  
  mod_label  
)
```

Arguments

`output_dir` **character** the directory in which all reports are searched and the overview is saved

`strata_column` **character** name of a study variable to stratify the report by. It can be null

`segment_column` **character** name of a metadata attribute usable to split the report in sections of variables. It can be null

`strata_column_label` **character** the label of the variable used as `strata_column`

`subgroup` **character** optional, to define subgroups of cases

`mod_label` **list** `util_ensure_label()` info

Value

an overview of all dataquieR reports created with dq_report_by

```
util_first_row_to_colnames
```

Move the first row of a data frame to its column names

Description

Move the first row of a data frame to its column names

Usage

```
util_first_row_to_colnames(dfr)
```

Arguments

dfr [data.frame](#)

Value

[data.frame](#) with first row as column names

util_get_encoding	<i>Get encoding from metadata or guess it from data</i>
-------------------	---

Description

Get encoding from metadata or guess it from data

Usage

```
util_get_encoding(  
  resp_vars = colnames(study_data),  
  study_data,  
  label_col,  
  meta_data,  
  meta_data_dataframe  
)
```

Arguments

resp_vars [variable](#) the names of the measurement variables, if missing or NULL, all variables will be checked

study_data [data.frame](#) the data frame that contains the measurements

label_col [variable attribute](#) the name of the column in the metadata with labels of variables

meta_data [data.frame](#) old name for item_level

meta_data_dataframe [data.frame](#) the data frame that contains the metadata for the data frame level

Value

named vector of valid encoding strings matching resp_vars

`util_has_no_group_vars`*Utility function to check whether a variable has no grouping variable assigned*

Description

Utility function to check whether a variable has no grouping variable assigned

Usage

```
util_has_no_group_vars(resp_vars, label_col = LABEL, meta_data = "item_level")
```

Arguments

`resp_vars` [variable list](#) the name of a measurement variable
`label_col` [variable attribute](#) the name of the column in the metadata with labels of variables
`meta_data` [data.frame](#) old name for `item_level`

Value

boolean

`util_histogram`*Utility function to create histograms*

Description

A helper function for simple histograms.

Usage

```
util_histogram(  
  plot_data,  
  num_var = colnames(plot_data)[1],  
  fill_var = NULL,  
  facet_var = NULL,  
  nbins_max = 100,  
  colors = "#2166AC",  
  is_datetime = FALSE  
)
```


Arguments

plot_data	a data.frame without missing values
num_var	column name of the numerical or datetime variable in plot_data (if omitted, the first column is assumed to contain this variable)
fill_var	column name of the categorical variable in plot_data which will be used for coloring stacked histograms
facet_var	column name of the categorical variable in plot_data which will be used to create facets
nbins_max	the maximum number of bins for the histogram (see util_optimize_histogram_bins)
colors	vector of colors, or a single color
is_datetime	if TRUE, the x-axis will be adapted for the datetime format

Value

a histogram

util_margins_bin	<i>Utility function to create a margins plot for binary variables</i>
------------------	---

Description

Utility function to create a margins plot for binary variables

Usage

```
util_margins_bin(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  threshold_type = NULL,
  threshold_value,
  min_obs_in_subgroup = 5,
  min_obs_in_cat = 5,
  caption = NULL,
  ds1,
  label_col,
  adjusted_hint = "",
  title = "",
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",
    dataquieR.acc_margins_sort_default),
  include_numbers_in_figures = getOption("dataquieR.acc_margins_num",
    dataquieR.acc_margins_num_default)
)
```

Arguments

resp_vars	variable the name of the binary measurement variable
group_vars	variable the name of the observer, device or reader variable
co_vars	variable list a vector of covariables, e.g. age and sex for adjustment
threshold_type	enum empirical user none. See acc_margins.
threshold_value	numeric see acc_margins
min_obs_in_subgroup	integer from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis.
min_obs_in_cat	integer This optional argument specifies the minimum number of observations that is required to include a category (level) of the outcome (resp_vars) in the analysis.
caption	string a caption for the plot (optional, typically used to report the coding of cases and control group)
ds1	data.frame the data frame that contains the measurements, after replacing missing value codes by NA, excluding inadmissible values and transforming categorical variables to factors.
label_col	variable attribute the name of the column in the metadata with labels of variables
adjusted_hint	character hint, if adjusted for co_vars
title	character title for the plot
sort_group_var_levels	logical Should the levels of the grouping variable be sorted descending by the number of observations (in the figure)?
include_numbers_in_figures	logical Should the figure report the number of observations for each level of the grouping variable?

Value

A table and a matching plot.

util_margins_lm	<i>Utility function to create a margins plot from linear regression models</i>
-----------------	--

Description

Utility function to create a margins plot from linear regression models

Usage

```

util_margins_lm(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  threshold_type = NULL,
  threshold_value,
  min_obs_in_subgroup = 5,
  ds1,
  label_col,
  levels = NULL,
  adjusted_hint = "",
  title = "",
  n_violin_max = getOption("dataquieR.max_group_var_levels_with_violins",
    dataquieR.max_group_var_levels_with_violins_default),
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",
    dataquieR.acc_margins_sort_default),
  include_numbers_in_figures = getOption("dataquieR.acc_margins_num",
    dataquieR.acc_margins_num_default)
)

```

Arguments

resp_vars [variable](#) the name of the measurement variable
group_vars [variable](#) the name of the observer, device or reader variable
co_vars [variable list](#) a vector of covariables, e.g. age and sex for adjustment
threshold_type [enum](#) empirical | user | none. See acc_margins.
threshold_value [numeric](#) see acc_margins
min_obs_in_subgroup [integer](#) from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis.
ds1 [data.frame](#) the data frame that contains the measurements, after replacing missing value codes by NA, excluding inadmissible values and transforming categorical variables to factors.
label_col [variable attribute](#) the name of the column in the metadata with labels of variables
levels [levels\(\)](#) of the original ordinal variable, if applicable. Used for axis tick labels.
adjusted_hint [character](#) hint, if adjusted for co_vars
title [character](#) title for the plot
n_violin_max [integer](#) from=0. This optional argument specifies the maximum number of levels of the group_var for which violin plots will be shown in the figure.
sort_group_var_levels [logical](#) Should the levels of the grouping variable be sorted descending by the number of observations (in the figure)?

include_numbers_in_figures

logical Should the figure report the number of observations for each level of the grouping variable?

Value

A table and a matching plot.

util_margins_nom	<i>Utility function to create a plot similar to the margins plots for nominal variables</i>
------------------	---

Description

This function is still under development. It uses the nnet package to compute multinomial logistic regression models.

Usage

```
util_margins_nom(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  min_obs_in_subgroup = 5,
  min_obs_in_cat = 5,
  ds1,
  label_col,
  adjusted_hint = "",
  title = "",
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",
    dataquieR.acc_margins_sort_default)
)
```

Arguments

resp_vars **variable** the name of the nominal measurement variable

group_vars **variable** the name of the observer, device or reader variable

co_vars **variable list** a vector of covariables, e.g. age and sex for adjustment

min_obs_in_subgroup **integer** from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis.

min_obs_in_cat **integer** This optional argument specifies the minimum number of observations that is required to include a category (level) of the outcome (resp_vars) in the analysis.

ds1	data.frame the data frame that contains the measurements, after replacing missing value codes by NA, excluding inadmissible values and transforming categorical variables to factors.
label_col	variable attribute the name of the column in the metadata with labels of variables
adjusted_hint	character hint, if adjusted for co_vars
title	character title for the plot
sort_group_var_levels	logical Should the levels of the grouping variable be sorted descending by the number of observations (in the figure)?

Value

A table and a matching plot.

util_margins_ord	<i>Utility function to create a plot similar to the margins plots for ordinal variables</i>
------------------	---

Description

This function is still under development. It uses the `ordinal` package to compute ordered regression models.

Usage

```
util_margins_ord(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  min_obs_in_subgroup = 5,
  min_subgroups = 5,
  ds1,
  label_col,
  adjusted_hint = "",
  title = "",
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",
    dataquieR.acc_margins_sort_default)
)
```

Arguments

resp_vars	variable the name of the ordinal measurement variable
group_vars	variable the name of the observer, device or reader variable
co_vars	variable list a vector of covariables, e.g. age and sex for adjustment

min_obs_in_subgroup	integer from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis.
min_subgroups	integer from=3. The model provided by the ordinal package requires at least three different subgroups (levels) of the group_var. Users might want to increase this threshold to obtain results only for variables with a sufficient number of group_var levels (observers, devices, etc.).
ds1	data.frame the data frame that contains the measurements, after replacing missing value codes by NA, excluding inadmissible values and transforming categorical variables to factors.
label_col	variable attribute the name of the column in the metadata with labels of variables
adjusted_hint	character hint, if adjusted for co_vars
title	character title for the plot
sort_group_var_levels	logical Should the levels of the grouping variable be sorted descending by the number of observations (in the figure)?

Value

A table and a matching plot.

util_margins_poi	<i>Utility function to create a margins plot from Poisson regression models</i>
------------------	---

Description

Utility function to create a margins plot from Poisson regression models

Usage

```
util_margins_poi(
  resp_vars = NULL,
  group_vars = NULL,
  co_vars = NULL,
  threshold_type = NULL,
  threshold_value,
  min_obs_in_subgroup = 5,
  ds1,
  label_col,
  adjusted_hint = "",
  title = "",
  sort_group_var_levels = getOption("dataquieR.acc_margins_sort",
    dataquieR.acc_margins_sort_default),
  include_numbers_in_figures = getOption("dataquieR.acc_margins_num",
    dataquieR.acc_margins_num_default)
)
```

Arguments

resp_vars	variable	the name of the measurement variable
group_vars	variable	the name of the observer, device or reader variable
co_vars	variable list	a vector of covariables, e.g. age and sex for adjustment
threshold_type	enum	empirical user none. See acc_margins.
threshold_value	numeric	see acc_margins
min_obs_in_subgroup	integer	from=0. This optional argument specifies the minimum number of observations that is required to include a subgroup (level) of the group_var in the analysis.
ds1	data.frame	the data frame that contains the measurements, after replacing missing value codes by NA, excluding inadmissible values and transforming categorical variables to factors.
label_col	variable attribute	the name of the column in the metadata with labels of variables
adjusted_hint	character	hint, if adjusted for co_vars
title	character	title for the plot
sort_group_var_levels	logical	Should the levels of the grouping variable be sorted descending by the number of observations (in the figure)?
include_numbers_in_figures	logical	Should the figure report the number of observations for each level of the grouping variable?

Value

A table and a matching plot.

util_plot_categorical_vars

Utility function to create plots for categorical variables

Description

Depending on the required level of complexity, this helper function creates various plots for categorical variables. Next to basic bar plots, it also enables group comparisons (for example for device/examiner effects) and longitudinal views.

Usage

```
util_plot_categorical_vars(  
  resp_vars,  
  group_vars = NULL,  
  time_vars = NULL,  
  study_data,  
  meta_data,  
  n_cat_max = 6,  
  n_group_max = getOption("dataquieR.max_group_var_levels_in_plot", 20),  
  n_data_min = 20  
)
```

Arguments

resp_vars	name of the categorical variable
group_vars	name of the grouping variable
time_vars	name of the time variable
study_data	the data frame that contains the measurements
meta_data	the data frame that contains metadata attributes of study data
n_cat_max	maximum number of categories to be displayed individually for the categorical variable (resp_vars)
n_group_max	maximum number of categories to be displayed individually for the grouping variable (group_vars, devices / examiners)
n_data_min	minimum number of data points to create a time course plot for an individual category of the resp_vars variable

Value

a figure

util_varcomp_robust *Utility function to compute the rank intraclass correlation*

Description

This implementation uses the package rankICC to compute the rank intraclass correlation, a non-parametric version of the ICC (Tu et al., 2023). In contrast to model-based ICC approaches, it is less sensitive to outliers and skewed distributions. It can be applied to variables with an ordinal, interval or ratio scale. However, it is not possible to adjust for covariables with this approach. The calculated ICC can become negative, like Fisher's ICC.

Usage

```
util_varcomp_robust(
  resp_vars = NULL,
  group_vars = NULL,
  study_data = study_data,
  meta_data = meta_data,
  min_obs_in_subgroup = 10,
  min_subgroups = 5,
  label_col = NULL
)
```

Arguments

<code>resp_vars</code>	the name of the response variable
<code>group_vars</code>	the name of the grouping variable
<code>study_data</code>	the data frame that contains the measurements
<code>meta_data</code>	the data frame that contains metadata attributes of study data
<code>min_obs_in_subgroup</code>	the minimum number of observations that is required to include a subgroup (level) of the grouping variable (<code>group_vars</code>) in the analysis. Subgroups with fewer observations are excluded.
<code>min_subgroups</code>	the minimum number of subgroups (levels) of the grouping variable (<code>group_vars</code>). If the variable has fewer subgroups, the analysis is not performed.
<code>label_col</code>	the name of the column in the metadata with labels of variables

Value

a vector from `rankICC::rankICC`

<code>value/missing-lists</code>	<i>Data frame with labels for missing- and jump-codes #' Metadata about value and missing codes</i>
----------------------------------	---

Description

`data.frame` with the following columns:

- `CODE_VALUE`: `numeric` | `DATETIME` Missing or categorical code (the number or date representing a missing/category)
- `CODE_LABEL`: `character` a label for the missing code or category
- `CODE_CLASS`: `enum` JUMP | MISSING. For missing lists: Class of the missing code.
- `CODE_INTERPRET` `enum` I | P | PL | R | BO | NC | O | UH | UO | NE. For missing lists: Class of the missing code according to [AAPOR](#).
- `resp_vars`: `character` For missing lists: optional, if a missing code is specific for some variables, it is listed for each such variable with one entry in `resp_vars`. If NA, the code is assumed shared among all variables. For v1.0 metadata, you need to refer to `VAR_NAMES` here.

See Also

[Online](#)

[com_item_missingness\(\)](#)

[com_segment_missingness\(\)](#)

[com_qualified_item_missingness\(\)](#)

[com_qualified_segment_missingness\(\)](#)

[con_inadmissible_categorical\(\)](#)

[con_inadmissible_vocabulary\(\)](#)

[MISSING_LIST_TABLE](#)

[VALUE_LABEL_TABLE](#)

[STANDARDIZED_VOCABULARY_TABLE](#)

[cause_label_df](#)

VARATT_REQUIRE_LEVELS *Requirement levels of certain metadata columns*

Description

These levels are cumulatively used by the function [prep_create_meta](#) and related in the argument `level` therein.

Usage

```
VARATT_REQUIRE_LEVELS
```

Format

An object of class `list` of length 5.

Details

currently available:

- `'COMPATIBILITY'` = "compatibility"
- `'REQUIRED'` = "required"
- `'RECOMMENDED'` = "recommended"
- `'OPTIONAL'` = "optional"
- `'TECHNICAL'` = "technical"

VARIABLE_LIST	<i>Cross-item level metadata attribute name</i>
---------------	---

Description

Specifies a group of variables for multivariate analyses. Separated by |, please use variable names from [VAR_NAMES](#) or a label as specified in `label_col`, usually [LABEL](#) or [LONG_LABEL](#).

Usage

VARIABLE_LIST

Format

An object of class character of length 1.

Details

if missing, `dataquieR` will create such IDs from [CONTRADICTION_TERM](#), if specified.

See Also

[meta_data_cross](#)

Other `meta_data_cross`: [ASSOCIATION_DIRECTION](#), [ASSOCIATION_FORM](#), [ASSOCIATION_METRIC](#), [ASSOCIATION_RANGE](#), [CHECK_ID](#), [CHECK_LABEL](#), [CONTRADICTION_TERM](#), [CONTRADICTION_TYPE](#), [DATA_PREPARATION](#), [GOLDSTANDARD](#), [MULTIVARIATE_OUTLIER_CHECK](#), [MULTIVARIATE_OUTLIER_CHECKTYPE](#), [N_RULES](#), [REL_VAL](#), [meta_data_cross](#), [util_normalize_cross_item\(\)](#)

VARIABLE_ROLES	<i>Variable roles can be one of the following:</i>
----------------	--

Description

- `intro` a variable holding consent-data
- `primary` a primary outcome variable
- `secondary` a secondary outcome variable
- `process` a variable describing the measurement process
- `suppress` a variable added on the fly computing sub-reports, i.e., by [dq_report_by](#) to have all referred variables available, even if they are not part of the currently processed segment. But they will only be fully assessed in their real segment's report.

Usage

VARIABLE_ROLES

Format

An object of class `list` of length 5.

WELL_KNOWN_META_VARIABLE_NAMES

Well-known metadata column names, names of metadata columns

Description

names of the variable attributes in the metadata frame holding the names of the respective observers, devices, lower limits for plausible values, upper limits for plausible values, lower limits for allowed values, upper limits for allowed values, the variable name (column name, e.g. v0020349) used in the study data, the variable name used for processing (readable name, e.g. RR_DIAST_1) and in parameters of the QA-Functions, the variable label, variable long label, variable short label, variable data type (see also [DATA_TYPES](#)), re-code for definition of lists of event categories, missing lists and jump lists as CSV strings. For valid units see [UNITS](#).

Usage

WELL_KNOWN_META_VARIABLE_NAMES

Format

An object of class `list` of length 58.

Details

all entries of this list will be mapped to the package's exported `NAMESPACE` environment directly, i.e. they are available directly by their names too:

- [VAR_NAMES](#)
- [LABEL](#)
- [DATA_TYPE](#)
- [SCALE_LEVEL](#)
- [UNIT](#)
- [VALUE_LABELS](#)
- [VALUE_LABEL_TABLE](#)
- [MISSING_LIST](#)
- [JUMP_LIST](#)
- [MISSING_LIST_TABLE](#)
- [HARD_LIMITS](#)
- [DETECTION_LIMITS](#)
- [SOFT_LIMITS](#)

- CONTRADICTIONS
- DISTRIBUTION
- DECIMALS
- DATA_ENTRY_TYPE
- END_DIGIT_CHECK
- CO_VARS
- GROUP_VAR_OBSERVER
- GROUP_VAR_DEVICE
- KEY_OBSERVER
- KEY_DEVICE
- TIME_VAR
- KEY_DATETIME
- PART_VAR
- STUDY_SEGMENT
- KEY_STUDY_SEGMENT
- VARIABLE_ROLE
- VARIABLE_ORDER
- LONG_LABEL
- SOFT_LIMIT_LOW
- SOFT_LIMIT_UP
- HARD_LIMIT_LOW
- HARD_LIMIT_UP
- DETECTION_LIMIT_LOW
- DETECTION_LIMIT_UP
- INCL_SOFT_LIMIT_LOW
- INCL_SOFT_LIMIT_UP
- INCL_HARD_LIMIT_LOW
- INCL_HARD_LIMIT_UP
- LOCATION_RANGE
- LOCATION_METRIC
- PROPORTION_RANGE
- LOCATION_LIMIT_LOW
- LOCATION_LIMIT_UP
- INCL_LOCATION_LIMIT_LOW
- INCL_LOCATION_LIMIT_UP
- PROPORTION_LIMIT_LOW
- PROPORTION_LIMIT_UP

- [INCL_PROPORTION_LIMIT_LOW](#)
- [INCL_PROPORTION_LIMIT_UP](#)
- [RECODE_CASES](#)
- [RECODE_CONTROL](#)
- [GRADING_RULESET](#)
- [STANDARDIZED_VOCABULARY_TABLE](#)
- [DATAFRAMES](#)
- [ENCODING](#)

See Also

[meta_data_segment](#) for STUDY_SEGMENT

Other UNITS: [UNITS](#), [UNIT_IS_COUNT](#), [UNIT_PREFIXES](#), [UNIT_SOURCES](#)

Examples

```
print(WELL_KNOWN_META_VARIABLE_NAMES$VAR_NAMES)
# print(VAR_NAMES) # should usually also work
```

```
[.dataquieR_resultset2
```

Get a subset of a dataquieR dq_report2 report

Description

Get a subset of a dataquieR dq_report2 report

Usage

```
## S3 method for class 'dataquieR_resultset2'
x[row, col, res, drop = FALSE, els = row]
```

Arguments

x	the report
row	the variable names, must be unique
col	the function-call-names, must be unique
res	the result slot, must be unique
drop	drop, if length is 1
els	used, if in list-mode with named argument

Value

a list with results, depending on drop and the number of results, the list may contain all requested results in sub-lists. The order of the results follows the order of the row/column/result-names given

```
[<-.dataquieR_resultset2  
      Write to a report
```

Description

Overwriting of elements only list-wise supported

Usage

```
## S3 replacement method for class 'dataquieR_resultset2'  
x[...] <- value
```

Arguments

<code>x</code>	a 'dataquieR_resultset2
<code>...</code>	if this contains only one entry and this entry is not named or its name is <code>els</code> , then, the report will be accessed in list mode.
<code>value</code>	new value to write

Value

nothing, stops

```
[ [.dataquieR_resultset2  
      Get a single result from a dataquieR 2 report
```

Description

Get a single result from a dataquieR 2 report

Usage

```
## S3 method for class 'dataquieR_resultset2'  
x[[e1]]
```

Arguments

<code>x</code>	the report
<code>e1</code>	the index

Value

the dataquieR result object

```
[[<- .dataquieR_resultset2
      Set a single result from a dataquieR 2 report
```

Description

Set a single result from a dataquieR 2 report

Usage

```
## S3 replacement method for class 'dataquieR_resultset2'
x[[e1]] <- value
```

Arguments

x	the report
e1	the index
value	the single result

Value

the dataquieR result object

```
$.dataquieR_resultset2
      Access single results from a dataquieR\_resultset2 report
```

Description

Access single results from a [dataquieR_resultset2](#) report

Usage

```
## S3 method for class 'dataquieR_resultset2'
x$e1
```

Arguments

x	the report
e1	the index

Value

the dataquieR result object

`$<-.dataquieR_resultset2`*Write single results from a `dataquieR_resultset2` report*

Description

Write single results from a `dataquieR_resultset2` report

Usage

```
## S3 replacement method for class 'dataquieR_resultset2'  
x$el <- value
```

Arguments

<code>x</code>	the report
<code>el</code>	the index
<code>value</code>	the single result

Value

the `dataquieR` result object

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