Package 'dataquieR'

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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> 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

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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 con 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:

- 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

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 con 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:

- 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 con 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:

- `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 con 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:

- 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` *Extension of acc_shape_or_scale to examine uniform distributions of end digits*

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` *Smoothes and plots adjusted longitudinal measurements and longitudinal trends from logistic regression models*

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 ggplot2::geom_line().
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 reordered.
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 a 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 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 con 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 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` *Utility function to compute model-based ICC depending on the (statistical) data type*

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 *Cross-item level metadata attribute name*

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 *Cross-item level metadata attribute name*

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 *Cross-item level metadata attribute name*

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 *Cross-item level metadata attribute name*

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 *Cross-item level metadata attribute name*

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_CH MULTIVARIATE_OUTLIER_CHECKTYPE, N_RULES, REL_VAL, VARIABLE_LIST, meta_data_cross, util_normalize_cross_it

---

check_table *Data frame with contradiction rules*

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 *types of value codes*

Description

types of value codes

Usage

```
CODE_CLASSES
```

Format

An object of class `list` of length 3.

CODE_LIST_TABLE *Default Name of the Table featuring Code Lists*

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 *Only existence is checked, order not yet used*

Description

Only existence is checked, order not yet used

Usage

```
CODE_ORDER
```

Format

An object of class `character` of length 1.

com_item_missingness *Summarize missingness columnwise (in variable)*

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 be 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 *NULL* 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 *color_gradient_direction*.
`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 elsewise 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 con 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`.
`return_flagged_study_data`	logical return `FlaggedStudyData` in the result
`return_limit_categorical`	logical if TRUE return limit deviations also for categorical 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.
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_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_CALL 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_leve 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.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_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_CALL dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.min_bw, dataquieR.acc_loess.min_p 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_leve 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.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_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_CALL 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_observa 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_leve 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.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_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_CALL 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_leve
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.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_CALL
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_leve
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.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_CALL 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_leve 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

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`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_CALL 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 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_leve 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.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_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_CALL 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.s 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_leve 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.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_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_CALL 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_leve
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

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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_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_CALL 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_binaryrecodelimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

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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_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_CALL 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

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dataquieR.CONDITIONS_LEVEL_TRHESHOLD

Log Level

Description

TODO

See Also

Other options: dataquieR, dataquieR.CONDITIONS_WITH_STACKTRACE, dataquieR.ELEMENT_MISSMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LAB dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescap dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.ma 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.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_MISSMATCH_CHECKTYPE, dataquieR.ERRORS_WITH_CALLER, dataquieR.GAM_for_LOESS, dataquieR.MAX_LABEL_LEN, dataquieR.MAX_VALUE_LAB dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescap dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.ma 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_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_CALL 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

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`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_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_CALL, 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_leve, 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

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`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_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_CALL, 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_binaryrecodelimit, dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

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dataquieR.ELEMENT_MISSMATCH_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_LAB, dataquieR.MESSAGES_WITH_CALLER, dataquieR.MULTIVARIATE_OUTLIER_CHECK, dataquieR.VALUE_LABELS_htmlescap, dataquieR.WARNINGS_WITH_CALLER, dataquieR.acc_loess.exclude_constant_subgroups, dataquieR.acc_loess.ma, 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

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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_MISSMATCH_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_consta 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_leve 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

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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 ore 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_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_CALL 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.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

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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_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_CALL 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.force_item_specific_missing_codes, dataquieR.force_label_ 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

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`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_CALL 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_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

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`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_CALL 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.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

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`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_MISSMATCH_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_consta, 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_leve, 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

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`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_CALL 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_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

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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_CALL 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.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.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_CALL 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.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.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_CALL 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.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

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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_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_CALL
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_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

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dataquieR.max_group_var_levels_with_violins

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

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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_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_CALL 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.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

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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_MISSMATCH_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_consta 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_leve 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.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_consta 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_leve 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_consta 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_leve
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.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_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_CALL, 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_binaryrecodelimit, 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_MISSMATCH_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_consta 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_leve 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_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_CALL 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_CALL 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_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_CALL 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.scale_level_heuristics_control_binaryreco dataquieR.scale_level_heuristics_control_metriclevels, dataquieR.testdebug

---

dataquieR.scale_level_heuristics_control_binaryrecodelimit

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_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_CALL 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.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_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_CALL 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.testdebug

---

`dataquieR.testdebug` *Disable all interactively used metadata-based function argument provision*

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_CALL 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.VALUE_LABELS_htmlescaped
Assume, all VALUE_LABELS are R*hrefhttps://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_cons 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_leve 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 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_leve 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_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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Data frame level metadata attribute name*

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 *Get the dimensions of a* `dq_report2` *result*

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 *Names of DQ dimensions*

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 *Names of a* `dataquieR` *report object (v2.0)*

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 *Generate a full DQ report*

Description

Deprecated

Usage

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

Arguments

...	Deprecated

Value

Deprecated

dq_report2 *Generate a full DQ report, v2*

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 parallelMap::parallelStart or NULL, 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 `options()`

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 `step` 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 `dq_report_by` (otherwise NULL)

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!

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

.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., bia), a data frame (e.g., "C:/Users/data/bia.dta"), a vector containing data frames files (e.g., c("C:/Users/data/bia.dta", C:/Users/data/biames.dta")), 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., "bia.dta"), the file name needs the extension and the path must be provided in the argument input_dir. It can also contain only the file name in case of example data from the package dataquieR (e.g., "study_data" or "ship")
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 study_data 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: c("missing_table1", "missing_table2")
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 meta_data_v2
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 segment_column nor strata_column or subgroup are defined. To create an un-split report please write explicitly the argument 'segment_column = NULL'
strata_column	variable name of a study variable to stratify the report by, e.g. the study centers. Both labels and VAR_NAMES 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 strata_column are limited to the content of this vector. A character vector or a regular expression can be provided (e.g., "^a.*$"). This argument can not be used if no strata_column is provided
selection_type	character optional, can only be specified if a strata_select or strata_exclude is specified. If not present the function try to guess what the user typed as strata_select or strata_exclude. There are 3 options: value indicating that the stratum selected is a value and not a value_label. For example "0"; v_label indicating that the stratum specified is a label. For example "male". regex indicating that the user specified strata using a regular expression. For example "^Ber" 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_CHE 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 missmatch 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_MISSMATCH_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 missmatch 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_MISSMATCH_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 duplicates 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` *Check for duplicated IDs*

Description

This function tests for duplicates 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_MISSMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISSMATCH_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_MISSMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISSMATCH_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_MISSMATCH_CHECKTYPE = "none")
int_sts_element_segment(study_data, meta_data)
options(dataquieR.ELEMENT_MISSMATCH_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 *Data frame with metadata about the study data on variable level*

Description

Variable level metadata.

See Also

further details on variable level metadata.

meta_data_segment

meta_data_dataframe

meta_data_cross *Well known columns on the* meta_data_cross-item *sheet*

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 *Well known columns on the* meta_data_dataframe *sheet*

Description

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

meta_data_segment *Well known columns on the* meta_data_segment *sheet*

Description

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

MULTIVARIATE_OUTLIER_CHECK *Cross-item level metadata attribute name*

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 *Cross-item and item level metadata attribute name*

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` *Call (nearly) one "Accuracy" function with many parameterizations at once automatically*

Description

Deprecated

Usage

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

Arguments

`...`	Deprecated

Value

Deprecated

`plot.dataquieR_summary` *Plot a* `dataquieR` *summary*

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 to 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 NA*s 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

study_data	data.frame the data frame that contains the measurements
meta_data	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: • VAR_NAMES: VAR_NAMES of the variable to compute • RULE: 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
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.

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 NA*s 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: • 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 formals still.

Value

a data frame with amended metadata.

---

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

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$DF_RECORD_COUNT <- NULL
print(prep_check_meta_data_dataframe(mds1)) # fixes the missing column by NAs
mds1 <- mds
mds1$DF_UNIQUE_ROWS[[2]] <- "xxx" # not convertible
# print(prep_check_meta_data_dataframe(mds1)) # fail
mds1 <- mds
mds1$DF_UNIQUE_ID[[2]] <- 12
# print(prep_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$SEGMENT_RECORD_COUNT <- NULL
print(prep_check_meta_data_segment(mds1)) # fixes the missing column by NAs
mds1 <- mds
mds1$SEGMENT_UNIQUE_ROWS[[2]] <- "xxx" # not convertible
# print(prep_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(prep_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_plotl`
`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 *Expand code labels across variables*

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.da
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_plotl 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_summar 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_summarycla 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(prep_get_data_frame(
  paste0("https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus",
    "/Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=",
    "publicationFile|COVID_Todesfälle_BL|Bundesland"))[[1]])

n <- as.numeric(prep_get_data_frame(paste0(
  "https://www.rki.de/DE/Content/InfAZ/N/Neuartiges_Coronavirus/",
  "Projekte_RKI/COVID-19_Todesfaelle.xlsx?__blob=",
  "publicationFile|COVID_Todesfälle_BL|Anzahl verstorbene",
  " COVID-19 Fälle"))[[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
#
prep_get_data_frame("meta_data | meta_data")

## End(Not run)
```

---

prep_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 *Prepare a label as part of a link for* RMD *files*

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 *List Loaded Data Frames*

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 *All valid* voc: *vocabularies*

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

folder	the folder name to load.
keep_types	logical keep types as possibly defined in the file. set TRUE for study data.
...	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(prep_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

---

prep_min_obs_level *Support function to identify the levels of a process variable with minimum number of observations*

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

.study_data	if provided, use this data set as study_data
.meta_data	if provided, use this data set as meta_data
.label_col	if provided, use this as label_col
.replace_hard_limits	replace HARD_LIMIT violations by NA, defaults to FALSE.
.replace_missings	replace missing codes, defaults to TRUE
.sm_code	missing code for NAs, if they have been re-coded by util_combine_missing_lists
.allow_empty	allow ds1 to be empty, i.e., 0 rows and/or 0 columns
.adjust_data_type	ensure that the data type of variables in the study data corresponds to their data type specified in the metadata
.amend_scale_level	ensure that SCALE_LEVEL is available in the item-level meta_data. internally used to prevent recursion, if called from prep_scalelevel_from_data_and_metadata().
.apply_factor_metadata	logical convert categorical variables to labeled factors.
.apply_factor_metadata_inadm	logical convert categorical variables to labeled factors keeping inadmissible values. Implies, that .apply_factor_metadata will be set to TRUE, too.
.internal	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_resul
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 prep_summary_to_classes but summarized by one column (currently, we support indicator_metric, call_names, STUDY_SEGMENT, and VAR_NAMES)
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_resul
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 floor(length(x) / (it/20))), 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` *Heuristics to amend a SCALE_LEVEL column and a UNIT column in the metadata*

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 storr storage.

Usage

```
prep_set_backend(r, storr_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 *Guess a metadata data frame from study data.*

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_resul`
`prep_render_pie_chart_from_summaryclasses_ggplot2()`, `prep_render_pie_chart_from_summaryclasses_plotl`
`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

x	the object
...	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

x	interval objects to print
...	not used yet

Value

the printed object

See Also

base::print

print.list *print a list of* dataquieR_result *objects*

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 *Print a* master_result *object*

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 con 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 *Print a* Slot *object*

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 *Print a* StudyDataSlot *object*

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 con 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_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`
segment_level	data.frame alias for `meta_data_segment`
dataframe_level	data.frame alias for `meta_data_dataframe`

Details

This is a preparatory support function that compares study data with associated metadata. A pre-requisite 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 *Return names of result slots (e.g., 3rd dimension of dataquieR results)*

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 *Return names of result slots (e.g., 3rd dimension of dataquieR results)*

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 *Scale Levels*

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 *Segment level metadata attribute name*

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 *Segment level metadata attribute name*

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 *Segment level metadata attribute name*

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 *Segment level metadata attribute name*

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 *Segment level metadata attribute name*

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 *Character used by default as a separator in metadata such as missing codes*

Description

This 1 character is according to our metadata concept "|".

Usage

```
SPLIT_CHAR
```

Format

An object of class `character` of length 1.

study_data *Data frame with the study data whose quality is being assessed*

Description

Study data is expected in wide format. If should contain all variables for all segments in one large table, even, if some variables are not measured for all observational utils (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 *Valid unit prefixes according to* `units::valid_udunits_prefixes()`

Description

like k, m, M, c, ...

See Also

Other UNITS: `UNITS`, `UNIT_IS_COUNT`, `UNIT_SOURCES`, `WELL_KNOWN_META_VARIABLE_NAMES`

UNIT_SOURCES *Maturity stage of a unit according to* `units::valid_udunits()`

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
Item level metadata attribute name

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

`plot_data`	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 (`fill_var`).
`cat_var`	column name of the categorical variable in `plot_data`
`num_var`	column name of the numerical variable in `plot_data`
`relative`	if `TRUE`, numbers will be interpreted as percentages (values in `num_var` should lie within `[0,1]`)
`show_numbers`	if `TRUE`, numbers will be displayed on top of the bars
`fill_var`	column name of the variable in `plot_data` which will be used to color the bars individually
`colors`	vector of colors, or a single color
`show_color_legend`	if `TRUE`, a legend for the colors will be displayed
`flip`	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 *Get encoding from metadata or guess it from data*

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 *Utility function to create a margins plot for binary variables*

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 *Utility function to create a margins plot from linear regression models*

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` *Utility function to create a plot similar to the margins plots for nominal variables*

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` *Utility function to create a plot similar to the margins plots for ordinal variables*

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 *Utility function to create a margins plot from Poisson regression models*

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

resp_vars	the name of the response variable
group_vars	the name of the grouping variable
study_data	the data frame that contains the measurements
meta_data	the data frame that contains metadata attributes of study data
min_obs_in_subgroup	the minimum number of observations that is required to include a subgroup (level) of the grouping variable (group_vars) in the analysis. Subgroups with fewer observations are excluded.
min_subgroups	the minimum number of subgroups (levels) of the grouping variable (group_vars). If the variable has fewer subgroups, the analysis is not performed.
label_col	the name of the column in the metadata with labels of variables

Value

a vector from rankICC::rankICC

---

value/missing-lists *Data frame with labels for missing- and jump-codes #' Metadata about value and missing codes*

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 *Cross-item level metadata attribute name*

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 *Variable roles can be one of the following:*

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

x	a 'dataquieR_resultset2
...	if this contains only one entry and this entry is not named or its name is els, then, the report will be accessed in list mode.
value	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[[el]]
```

Arguments

x	the report
el	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[[el]] <- value
```

Arguments

`x`	the report
`el`	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$el
```

Arguments

`x`	the report
`el`	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

`x`	the report
`el`	the index
`value`	the single result

Value

the `dataquieR` result object

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