Package ‘tern’

June 21, 2024

Title Create Common TLGs Used in Clinical Trials
Version 0.9.5
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Description Table, Listings, and Graphs (TLG) library for common outputs used in clinical trials.
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BugReports https://github.com/insightsengineering/tern/issues

Depends R (>= 3.6), rtables (>= 0.6.8)

Imports broom (>= 0.5.4), car (>= 3.0-13), checkmate (>= 2.1.0),
cowplot (>= 1.0.0), dplyr (>= 1.0.0), emmeans (>= 1.8.0),
forcats (>= 1.0.0), formatters (>= 0.5.8), ggplot2 (>= 3.5.0),
ggridExtra (>= 2.0.0), gtable (>= 0.3.0), labeling,
lifecycle (>= 0.2.0), magrittr (>= 1.5), MASS (>= 7.3-60),
methods, Rdpack (>= 2.4), rlang (>= 0.1.1), scales (>= 1.2.0),
stats, survival (>= 3.2-13), tibble (>= 2.0.0), tidyr (>= 0.8.3), utils

Suggests knitr (>= 1.42), lattice (>= 0.18-4), lubridate (>= 1.7.9),
nestcolor (>= 0.1.1), rmarkdown (>= 2.19), stringr (>= 1.4.1),
svglite (>= 2.1.2), testthat (>= 3.1.9), withr (>= 2.0.0)

VignetteBuilder knitr

RdMacros lifecycle, Rdpack

Config/Needs/verdepcheck insightsengineering/rtables,
tidymodels/broom, cran/car, mllg/checkmate, wilkelab/cowplot,
tidyverse/dplyr, rvlenth/emmeans, tidyverseforcats,
isightsengineering/formatters, tidyverse/ggplot2,
r-lib/gtable, r-lib/lifecycle, tidyverse/magrittr,
GeoBosh/Rdpack, r-lib/rlang, r-lib/scales, tidyverse/tibble,
tidyverse/tidyverse, yihui/knitr, deepayan/lattice,
tidyverse/lubridate, insightsengineering/nestcolor,
rstudio/rmarkdown, tidyverse/stringr, r-lib/svglite,
r-lib/testthat, r-lib/withr
Config/Needs/website: insightsengineering/nesttemplate

Config/testthat/edition: 3

Encoding: UTF-8
Language: en-US
LazyData: true
RoxygenNote: 7.3.1

Collate: 'formatting_functions.R' 'abnormal.R' 'abnormal_by_baseline.R'
'abnormal_by_marked.R' 'abnormal_by_worst_grade.R'
'abnormal_worst_grade_worsen.R'
'analyze_colvars_functions.R' 'analyze_functions.R'
'analyze_variables.R' 'analyze_vars_in_cols.R'
'argument_convention.R' 'bland_altman.R'
'combination_function.R' 'compare_variables.R'
'control_incidence_rate.R' 'control_logistic.R'
'control_step.R' 'control_survival.R' 'count_cumulative.R'
'count_missed_doses.R' 'count_occurrences.R'
'count_occurrences_by_grade.R'
'count_patients_events_in_cols.R' 'count_patients_with_event.R'
'count_patients_with_flags.R' 'count_values.R'
'cox_regression.R' 'cox_regression_inter.R' 'coxph.R'
'd_pkparam.R' 'data.R' 'decorate_grob.R'
destools_binom_diff.R' 'df_explicit_na.R'
estimate_multinomial_rsp.R' 'estimate_proportion.R'
'fit_rsp_step.R' 'fit_survival_step.R' 'g_forest.R' 'g_ipp.R'
'g_km.R' 'g_lineplot.R' 'g_step.R' 'g_waterfall.R'
'h_adsl_adlb_merge_using_worst_flag.R'
'h_biomarkers_subgroups.R' 'h_cox_regression.R' 'h_km.R'
'h_logistic_regression.R' 'h_map_for_count_abnormal.R'
'h_pkparam_sort.R' 'h_response_biomarkers_subgroups.R'
'h_response_subgroups.R' 'h_stack_by_baskets.R' 'h_step.R'
'h_survival_biomarkers_subgroups.R'
'h_survival_duration_subgroups.R' 'himputation_rule.R'
'incidence_rate.R' 'logistic_regression.R' 'missing_data.R'
'odds_ratio.R' 'package.R' 'prop_diff.R' 'prop_diff_test.R'
'prune_occurrences.R' 'response_biomarkers_subgroups.R'
'response_subgroups.R' 'riskdiff.R' 'rtables_access.R'
'score_occurrences.R' 'split_cols_by_groups.R' 'stat.R'
'summarize_ancova.R' 'summarize_change.R' 'summarize_colvars.R'
'summarize_coxreg.R' 'summarize_functions.R'
'summarize_glm_count.R' 'summarize_num_patients.R'
'summarize_patients_exposure_in_cols.R'
'summarize_biomarkers_subgroups.R' 'summarize_coxph_pairwise.R'
'summarize_duration_subgroups.R' 'survival_time.R'
'survival_timepoint.R' 'utils.R' 'utils_checkmate.R'
'utils_default_stats_formats_labels.R' 'utils_factor.R'
'utils_ggplot.R' 'utils_grid.R' 'utils_rtables.R'
'utils_split_funs.R'
Contents

NeedsCompilation no

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Contents

tern-package ......................................................... 7
add_riskdiff .......................................................... 8
add_rowcounts ......................................................... 9
aezi_label ............................................................. 10
analyze_colvars_functions .......................................... 11
analyze_functions .................................................... 11
analyze_variables .................................................... 13
analyze_vars_in_cols ............................................... 21
append_varlabels ..................................................... 25
arrange_grobs ......................................................... 27
as.rtable ............................................................. 28
bland_altman .......................................................... 29
combination_function ................................................ 30
combine_counts ......................................................... 32
combine_groups ......................................................... 33
combine_levels ......................................................... 34
combine_vectors ......................................................... 34
compare_variables ..................................................... 35
control_analyze_vars ............................................... 40
control_annot ........................................................ 41
control_coxph ........................................................ 42
control_coxreg ......................................................... 43
control_incidence_rate .............................................. 44
control_lineplot_vars ............................................... 45
control_logistic ....................................................... 46
control_step .......................................................... 47
<table>
<thead>
<tr>
<th>Function Name</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>control_surv_time</td>
<td>48</td>
</tr>
<tr>
<td>control_surv_timepoint</td>
<td>49</td>
</tr>
<tr>
<td>count_occurrences</td>
<td>49</td>
</tr>
<tr>
<td>count_occurrences_by_grade</td>
<td>54</td>
</tr>
<tr>
<td>count_patients_with_event</td>
<td>59</td>
</tr>
<tr>
<td>count_patients_with_flags</td>
<td>63</td>
</tr>
<tr>
<td>count_values_funs</td>
<td>67</td>
</tr>
<tr>
<td>cox_regression</td>
<td>70</td>
</tr>
<tr>
<td>cox_regression_inter</td>
<td>76</td>
</tr>
<tr>
<td>cut_quantile_bins</td>
<td>79</td>
</tr>
<tr>
<td>day2month</td>
<td>81</td>
</tr>
<tr>
<td>decorate_grob</td>
<td>82</td>
</tr>
<tr>
<td>decorate_grob_set</td>
<td>85</td>
</tr>
<tr>
<td>default_na_str</td>
<td>87</td>
</tr>
<tr>
<td>default_stats_formats_labels</td>
<td>88</td>
</tr>
<tr>
<td>df_explicit_na</td>
<td>91</td>
</tr>
<tr>
<td>draw_grob</td>
<td>93</td>
</tr>
<tr>
<td>d_count_abnormal_by_baseline</td>
<td>94</td>
</tr>
<tr>
<td>d_count_cumulative</td>
<td>95</td>
</tr>
<tr>
<td>d_count_missed_doses</td>
<td>95</td>
</tr>
<tr>
<td>d_onco_rsp_label</td>
<td>96</td>
</tr>
<tr>
<td>d_pkparam</td>
<td>97</td>
</tr>
<tr>
<td>d_proportion</td>
<td>97</td>
</tr>
<tr>
<td>d_proportion_diff</td>
<td>98</td>
</tr>
<tr>
<td>d_rsp_subgroups_colvars</td>
<td>98</td>
</tr>
<tr>
<td>d_survival_subgroups_colvars</td>
<td>99</td>
</tr>
<tr>
<td>d_test_proportion_diff</td>
<td>100</td>
</tr>
<tr>
<td>estimate_multinomial_rsp</td>
<td>101</td>
</tr>
<tr>
<td>estimate_proportions</td>
<td>103</td>
</tr>
<tr>
<td>explicit_na</td>
<td>107</td>
</tr>
<tr>
<td>extract_rsp_biomarkers</td>
<td>108</td>
</tr>
<tr>
<td>extract_rsp_subgroups</td>
<td>110</td>
</tr>
<tr>
<td>extract_survival_biomarkers</td>
<td>111</td>
</tr>
<tr>
<td>extract_survival_subgroups</td>
<td>112</td>
</tr>
<tr>
<td>extreme_format</td>
<td>113</td>
</tr>
<tr>
<td>ex_data</td>
<td>115</td>
</tr>
<tr>
<td>fctCollapse_only</td>
<td>116</td>
</tr>
<tr>
<td>fct_discard</td>
<td>117</td>
</tr>
<tr>
<td>fct_explicit_na_if</td>
<td>117</td>
</tr>
<tr>
<td>fit_coxreg</td>
<td>118</td>
</tr>
<tr>
<td>fit_logistic</td>
<td>121</td>
</tr>
<tr>
<td>fit_rsp_step</td>
<td>123</td>
</tr>
<tr>
<td>fit_survival_step</td>
<td>125</td>
</tr>
<tr>
<td>forest_viewport</td>
<td>127</td>
</tr>
<tr>
<td>formatting_functions</td>
<td>128</td>
</tr>
<tr>
<td>format_auto</td>
<td>129</td>
</tr>
<tr>
<td>format_count_fraction</td>
<td>130</td>
</tr>
<tr>
<td>format_count_fraction_fixed_dp</td>
<td>131</td>
</tr>
</tbody>
</table>
format_count_fraction_lt10 .................................................. 131
format_extreme_values ...................................................... 132
format_extreme_values_ci ................................................... 133
format_fraction .............................................................. 134
format_fraction_fixed_dp .................................................. 135
format_fraction_threshold ................................................ 136
format_sigfig ................................................................ 137
format_xx ................................................................ 138
f_conf_level ................................................................ 139
f_pval .................................................................. 139
get_smooths ................................................................ 140
groups_list_to_df ............................................................. 140
g_forest .................................................................. 141
g_ipp .................................................................... 145
g_km ..................................................................... 148
g_lineplot ................................................................. 153
g_step .................................................................. 158
g_waterfall ................................................................ 160
h_adlb_abnormal_by_worst_grade ..................................... 162
h_adlb_worsen ............................................................... 163
h_adsl_adlb_merge_using_worst_flag .................................. 165
h_ancova .................................................................. 166
h_append_grade_groups ..................................................... 167
h_col_indices ................................................................. 168
h_count_cumulative .......................................................... 169
h_cox_regression ............................................................. 170
h_data_plot ................................................................. 173
h_decompose_gg .............................................................. 175
h_format_row ................................................................. 176
h_ggkm .................................................................. 177
h_grob_coxph ................................................................. 179
h_grob_median_surv ........................................................ 180
h_grob_tbl_at_risk ........................................................... 181
h_grob_y_annot ............................................................... 183
h_g_ipp .................................................................. 184
h_km_layout ................................................................. 186
h_logistic_regression ........................................................ 188
h_map_for_count_abnormal .............................................. 192
h_odds_ratio ................................................................. 194
h_pkparam_sort ............................................................... 195
h_proportions ................................................................. 196
h_prop_diff ................................................................. 198
h_response_biomarkers_subgroups .................................... 201
h_response_subgroups ...................................................... 204
h_split_by_subgroups ...................................................... 207
h_split_param ............................................................... 209
h_stack_by_baskets .......................................................... 210
h_step ................................................................ 211
tern-package

Description

Package to create tables, listings and graphs to analyze clinical trials data.

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See Also

Useful links:

• https://insightsengineering.github.io/tern/
• https://github.com/insightsengineering/tern/
• Report bugs at https://github.com/insightsengineering/tern/issues
Description

[Stable]

Wrapper function for `rtables::add_combo_levels()` which configures settings for the risk difference column to be added to an `rtables` object. To add a risk difference column to a table, this function should be used as `split_fun` in calls to `rtables::split_cols_by()`, followed by setting argument `riskdiff` to `TRUE` in all following `analyze` function calls.

Usage

```r
add_riskdiff(
  arm_x,
  arm_y,
  col_label = paste0("Risk Difference (%) (95% CI)", if (length(arm_y) > 1)
               paste0("\n", arm_x, " vs. ", arm_y)),
  pct = TRUE
)
```

Arguments

- `arm_x` (string) name of reference arm to use in risk difference calculations.
- `arm_y` (character) names of one or more arms to compare to reference arm in risk difference calculations. A new column will be added for each value of `arm_y`.
- `col_label` (character) labels to use when rendering the risk difference column within the table. If more than one comparison arm is specified in `arm_y`, default labels will specify which two arms are being compared (reference arm vs. comparison arm).
- `pct` (flag) whether output should be returned as percentages. Defaults to `TRUE`.

Value

A closure suitable for use as a split function (`split_fun`) within `rtables::split_cols_by()` when creating a table layout.

See Also

`stat_propdiff_ci()` for details on risk difference calculation.
add_rowcounts

Examples

```r
daea <- tern_ex_adae
daee$AESEV <- factor(daea$AESEV)

lyt <- basic_table() %>%
split_cols_by("ARMCD", split_fun = add_riskdiff(arm_x = "ARM A", arm_y = c("ARM B", "ARM C"))) %>%
count_occurrences_by_grade(
  var = "AESEV",
  riskdiff = TRUE
)

tbl <- build_table(lyt, df = adae)
tbl
```

add_rowcounts  Layout-creating function to add row total counts

Description

[Stable]

This works analogously to rtables::add_colcounts() but on the rows. This function is a wrapper for rtables::summarize_row_groups().

Usage

```r
add_rowcounts(lyt, alt_counts = FALSE)
```

Arguments

- **lyt** (PreDataTableLayouts)
  layout that analyses will be added to.
- **alt_counts** (flag)
  whether row counts should be taken from alt_counts_df (TRUE) or from df (FALSE). Defaults to FALSE.

Value

A modified layout where the latest row split labels now have the row-wise total counts (i.e. without column-based subsetting) attached in parentheses.

Note

Row count values are contained in these row count rows but are not displayed so that they are not considered zero rows by default when pruning.
Examples

basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  split_rows_by("RACE", split_fun = drop_split_levels) %>%
  add_rowcounts() %>%
  analyze("AGE", afun = list_wrap_x(summary), format = "xx.xx") %>%
  build_table(DM)

aes_i_label

Labels for adverse event baskets

Description

[Stable]

Usage

aes_i_label(aesi, scope = NULL)

Arguments

aes_i (character)
  vector with standardized MedDRA query name (e.g. SMQxxNAM) or customized query name (e.g. CQxxNAM).

scope (character)
  vector with scope of query (e.g. SMQxxSC).

Value

A string with the standard label for the AE basket.

Examples

adae <- tern_ex_adae

# Standardized query label includes scope.
aesi_label(adae$SMQ01NAM, scope = adae$SMQ01SC)

# Customized query label.
aesi_label(adae$CQ01NAM)
analyze_colvars_functions

Analyze functions on columns

Description

These functions are wrappers of `rtables::analyze_colvars()` which apply corresponding tern statistics functions to add an analysis to a given table layout. In particular, these functions where designed to have the analysis methods split into different columns.

- **analyze_vars_in_cols()**: fundamental tabulation of analysis methods onto columns. In other words, the analysis methods are defined in the column space, i.e. they become column labels. By changing the variable vector, the list of functions can be applied on different variables, with the caveat of having the same number of statistical functions.

- **tabulate_rsp_subgroups()**: similarly to `analyze_vars_in_cols`, this function combines `analyze_colvars` and `summarize_row_groups` in a compact way to produce standard tables that show analysis methods as columns.

- **tabulate_survival_subgroups()**: this function is very similar to the above, but it is used for other tables.

- **analyze_patients_exposure_in_cols()**: based only on `analyze_colvars`. It needs `summarize_patients_exposure()` to leverage nesting of label rows analysis with `rtables::summarize_row_groups()`.

- **summarize_coxreg()**: generally based on `rtables::summarize_row_groups()`, it behaves similarly to `tabulate_*` functions described above as it is designed to provide specific standard tables that may contain nested structure with a combination of `summarize_row_groups()` and `rtables::analyze_colvars()`.

See Also

- `summarize_functions` for functions which are wrappers for `rtables::summarize_row_groups()`.
- `analyze_functions` for functions which are wrappers for `rtables::analyze()`.

analyze_functions

Analyze functions

Description

These functions are wrappers of `rtables::analyze()` which apply corresponding tern statistics functions to add an analysis to a given table layout:

- **analyze_num_patients()**
- **compare_vars()**
- **count_abnormal()**
- **count_abnormal_by_baseline()**
• count_abnormal_by_marked()
• count_abnormal_by_worst_grade()
• count_cumulative()
• count_missed_doses()
• count_occurrences()
• count_occurrences_by_grade()
• count_patients_events_in_cols()
• count_patients_with_event()
• count_patients_with_flags()
• count_values()
• coxph_pairwise()
• estimate_incidence_rate()
• estimate_multinomial_rsp()
• estimate_odds_ratio()
• estimate_proportion()
• estimate_proportion_diff()
• summarize_ancova()
• summarize_colvars(): even if this function uses rtables::analyze_colvars(), it applies the analysis methods as different rows for one or more variables that are split into different columns. In comparison, analyze_colvars_functions leverage analyze_colvars to have the context split in rows and the analysis methods in columns.
• summarize_change()
• analyze_vars()
• surv_time()
• surv_timepoint()
• test_proportion_diff()

See Also

• summarize_functions for functions which are wrappers for rtables::summarize_row_groups().
• analyze_colvars_functions for functions that are wrappers for rtables::analyze_colvars().
**analyze_variables**

**Analyze variables**

**Description**

**[Stable]**

The analyze function `analyze_vars()` generates a summary of one or more variables, using the S3 generic function `s_summary()` to calculate a list of summary statistics. A list of all available statistics for numeric variables can be viewed by running `get_stats("analyze_vars_numeric")` and for non-numeric variables by running `get_stats("analyze_vars_counts").` Use the `.stats` parameter to specify the statistics to include in your output summary table.

**Usage**

```r
analyze_vars(
  lyt,
  vars,
  var_labels = vars,
  na_str = default_na_str(),
  nested = TRUE,
  ..., 
  na.rm = TRUE,
  show_labels = "default",
  table_names = vars,
  section_div = NA_character_,
  .stats = c("n", "mean_sd", "median", "range", "count_fraction"),
  .formats = NULL,
  .labels = NULL,
  .indent_mods = NULL
)
```

```r
s_summary(x, na.rm = TRUE, denom, .N_row, .N_col, .var, ...)
```

```r
## S3 method for class 'numeric'

s_summary(
  x,
  na.rm = TRUE,
  denom,
  .N_row,
  .N_col,
  .var,
  control = control_analyze_vars(),
  ...
)
```

```r
## S3 method for class 'factor'
```
analyze_variables

s_summary(
  x,
  na.rm = TRUE,
  denom = c("n", "N_row", "N_col"),
  .N_row,
  .N_col,
  ...
)

## S3 method for class 'character'
s_summary(
  x,
  na.rm = TRUE,
  denom = c("n", "N_row", "N_col"),
  .N_row,
  .N_col,
  .var,
  verbose = TRUE,
  ...
)

## S3 method for class 'logical'
s_summary(
  x,
  na.rm = TRUE,
  denom = c("n", "N_row", "N_col"),
  .N_row,
  .N_col,
  ...
)

a_summary(
  x,
  .N_col,
  .N_row,
  .var = NULL,
  .df_row = NULL,
  .ref_group = NULL,
  .in_ref_col = FALSE,
  compare = FALSE,
  .stats = NULL,
  .formats = NULL,
  .labels = NULL,
  .indent_mods = NULL,
  na.rm = TRUE,
  na_str = default_na_str(),
  ...
)
analyze_variables

Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

vars (character)
variable names for the primary analysis variable to be iterated over.

var_labels (character)
variable labels.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.

... arguments passed to s_summary().

na.rm (flag)
whether NA values should be removed from x prior to analysis.

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

table_names (character)
this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

section_div (string)
string which should be repeated as a section divider after each group defined by this split instruction, or NA_character_ (the default) for no section divider.

.stats (character)
statistics to select for the table. Run get_stats("analyze_vars_numeric") to see statistics available for numeric variables, and get_stats("analyze_vars_counts") for statistics available for non-numeric variables.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Each element of the vector should be a name-value pair with name corresponding to a statistic specified in .stats and value the indentation for that statistic’s row label.

x (numeric)
vector of numbers we want to analyze.

denom (string)
choice of denominator for proportion. Options are:

- n: number of values in this row and column intersection.
- N_row: total number of values in this row across columns.
• N_col: total number of values in this column across rows.

.N_row  (integer(1))
row-wise N (row group count) for the group of observations being analyzed (i.e.
with no column-based subsetting) that is typically passed by rtables.

.N_col  (integer(1))
column-wise N (column count) for the full column being analyzed that is typi-
cally passed by rtables.

.var  (string)
single variable name that is passed by rtables when requested by a statistics
function.

control  (list)
parameters for descriptive statistics details, specified by using the helper func-
tion control_analyze_vars(). Some possible parameter options are:
  • conf_level (proportion)
    confidence level of the interval for mean and median.
  • quantiles (numeric(2))
    vector of length two to specify the quantiles.
  • quantile_type (numeric(1))
    between 1 and 9 selecting quantile algorithms to be used. See more about
type in stats::quantile().
  • test_mean (numeric(1))
    value to test against the mean under the null hypothesis when calculating
    p-value.

.verbose  (flag)
defaults to TRUE, which prints out warnings and messages. It is mainly used to
print out information about factor casting.

df_row  (data.frame)
data frame across all of the columns for the given row split.

.ref_group  (data.frame or vector)
the data corresponding to the reference group.

.in_ref_col  (flag)
TRUE when working with the reference level, FALSE otherwise.

.compare  (flag)
whether comparison statistics should be analyzed instead of summary statistics
(compare = TRUE adds pval statistic comparing against reference group).

Details

Automatic digit formatting: The number of digits to display can be automatically determined
from the analyzed variable(s) (vars) for certain statistics by setting the statistic format to "auto" in
.formats. This utilizes the format_auto() formatting function. Note that only data for the current
row & variable (for all columns) will be considered (.df_row[[.var]], see rtables::additional_fun_params)
and not the whole dataset.
analyze_variables

Value

- `analyze_vars()` returns a layout object suitable for passing to further layouting functions, or to `rtables::build_table()`. Adding this function to an rtable layout will add formatted rows containing the statistics from `s_summary()` to the table layout.

- `s_summary()` returns different statistics depending on the class of `x`.

  - If `x` is of class numeric, returns a list with the following named numeric items:
    - `n`: The `length()` of `x`.
    - `sum`: The `sum()` of `x`.
    - `mean`: The `mean()` of `x`.
    - `sd`: The `stats::sd()` of `x`.
    - `se`: The standard error of `x` mean, i.e.: `(sd(x) / sqrt(length(x)))`.
    - `mean_sd`: The mean and `stats::sd()` of `x`.
    - `mean_se`: The `mean()` of `x` and its standard error (see above).
    - `mean_ci`: The CI for the mean of `x` (from `stat_mean_ci()`).
    - `mean_sei`: The SE interval for the mean of `x` (from `mean() -/+ stats::sd() / sqrt()`).
    - `mean_sdi`: The SD interval for the mean of `x`, i.e.: `(mean() -/+ stats::sd())`.
    - `mean_pval`: The two-sided p-value of the mean of `x` (from `stat_mean_pval()`).
    - `median`: The `stats::median()` of `x`.
    - `mad`: The median absolute deviation of `x`, i.e.: `(stats::median() of xc, where xc = x - stats::median())`.
    - `median_ci`: The CI for the median of `x` (from `stat_median_ci()`).
    - `quantiles`: Two sample quantiles of `x` (from `stats::quantile()`).
    - `iqr`: The `stats::IQR()` of `x`.
    - `range`: The `range_noInf()` of `x`.
    - `min`: The `min()` of `x`.
    - `max`: The `max()` of `x`.
    - `median_range`: The `median()` and `range_noInf()` of `x`.
    - `cv`: The coefficient of variation of `x`, i.e.: `(stats::sd() / mean() * 100)`.
    - `geom_mean`: The geometric mean of `x`, i.e.: `(exp(mean(log(x)))`.
    - `geom_cv`: The geometric coefficient of variation of `x`, i.e.: `(sqrt(exp(sd(log(x)) ^ 2) - 1) * 100)`.

  - If `x` is of class factor or converted from character, returns a list with named numeric items:
    - `n`: The `length()` of `x`.
    - `count`: A list with the number of cases for each level of the factor `x`.
    - `count_fraction`: Similar to `count` but also includes the proportion of cases for each level of the factor `x` relative to the denominator, or `NA` if the denominator is zero.

  - If `x` is of class logical, returns a list with named numeric items:
    - `n`: The `length()` of `x` (possibly after removing NAs).
    - `count`: Count of `TRUE` in `x`.
- `count_fraction`: Count and proportion of TRUE in `x` relative to the denominator, or NA if the denominator is zero. Note that NAs in `x` are never counted or leading to NA here.

- `a_summary()` returns the corresponding list with formatted `rtables::CellValue()`.

### Functions

- `analyze_vars()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `s_summary()`: S3 generic function to produces a variable summary.
- `s_summary(numeric)`: Method for numeric class.
- `s_summary(factor)`: Method for factor class.
- `s_summary(character)`: Method for character class. This makes an automatic conversion to factor (with a warning) and then forwards to the method for factors.
- `s_summary(logical)`: Method for logical class.
- `a_summary()`: Formatted analysis function which is used as `afun` in `analyze_vars()` and `compare_vars()` and as `cfun` in `summarize_colvars()`.

### Note

- If `x` is an empty vector, NA is returned. This is the expected feature so as to return rcell content in `rtables` when the intersection of a column and a row delimits an empty data selection.

- When the `mean` function is applied to an empty vector, NA will be returned instead of NaN, the latter being standard behavior in R.

- If `x` is an empty factor, a list is still returned for counts with one element per factor level. If there are no levels in `x`, the function fails.

- If factor variables contain NA, these NA values are excluded by default. To include NA values set `na.rm = FALSE` and missing values will be displayed as an NA level. Alternatively, an explicit factor level can be defined for NA values during pre-processing via `df_explicit_na()` - the default `na_level` ("<Missing>") will also be excluded when `na.rm` is set to TRUE.

- Automatic conversion of character to factor does not guarantee that the table can be generated correctly. In particular for sparse tables this very likely can fail. It is therefore better to always pre-process the dataset such that factors are manually created from character variables before passing the dataset to `rtables::build_table()`.

- To use for comparison (with additional p-value statistic), parameter `compare` must be set to TRUE.

- Ensure that either all NA values are converted to an explicit NA level or all NA values are left as is.
## Fabricated dataset.
dta_test <- data.frame(
    USUBJID = rep(1:6, each = 3),
    PARAMCD = rep("lab", 6 * 3),
    AVISIT = rep(paste0("V", 1:3), 6),
    ARM = rep(LETTERS[1:3], rep(6, 3)),
    AVAL = c(9:1, rep(NA, 9))
)

# `analyze_vars()` in `rtables` pipelines
## Default output within a `rtables` pipeline.
l <- basic_table() %>%
    split_cols_by(var = "ARM") %>%
    split_rows_by(var = "AVISIT") %>%
    analyze_vars(vars = "AVAL")

build_table(l, df = dta_test)

## Select and format statistics output.
l <- basic_table() %>%
    split_cols_by(var = "ARM") %>%
    split_rows_by(var = "AVISIT") %>%
    analyze_vars(
        vars = "AVAL",
        .stats = c("n", "mean_sd", "quantiles"),
        .formats = c("mean_sd" = "xx.x, xx.x"),
        .labels = c(n = "n", mean_sd = "Mean, SD", quantiles = c("Q1 - Q3"))
    )

build_table(l, df = dta_test)

## Use arguments interpreted by `s_summary`.
l <- basic_table() %>%
    split_cols_by(var = "ARM") %>%
    split_rows_by(var = "AVISIT") %>%
    analyze_vars(vars = "AVAL", na.rm = FALSE)

build_table(l, df = dta_test)

## Handle `NA` levels first when summarizing factors.
dta_test$AVISIT <- NA_character_
dta_test <- df_explicit_na(dta_test)
l <- basic_table() %>%
    split_cols_by(var = "ARM") %>%
    analyze_vars(vars = "AVISIT", na.rm = FALSE)

build_table(l, df = dta_test)

# auto format
dt <- data.frame("VAR" = c(0.001, 0.2, 0.0011000, 3, 4))
basic_table() %>%
analyze_vars(
  vars = "VAR",
  .stats = c("n", "mean", "mean_sd", "range"),
  .formats = c("mean_sd" = "auto", "range" = "auto")
) %>%
build_table(dt)

# `s_summary.numeric`

## Basic usage: empty numeric returns NA-filled items.
s_summary(numeric())

## Management of NA values.
x <- c(NA_real_, 1)
s_summary(x, na.rm = TRUE)
s_summary(x, na.rm = FALSE)

x <- c(NA_real_, 1, 2)
s_summary(x, stats = NULL)

## Benefits in `rtables` contructions:
dta_test <- data.frame(
  Group = rep(LETTERS[1:3], each = 2),
  sub_group = rep(letters[1:2], each = 3),
  x = 1:6
)

## The summary obtained in with `rtables`:
基本_table() %>%
  split_cols_by(var = "Group") %>%
  split_rows_by(var = "sub_group") %>%
  analyze(vars = "x", afun = s_summary) %>%
  build_table(df = dta_test)

## By comparison with `lapply`:
X <- split(dta_test, f = with(dta_test, interaction(Group, sub_group)))
lapply(X, function(x) s_summary(x$x))

# `s_summary.factor`

## Basic usage:
s_summary(factor(c("a", "a", "b", "c", "a")))

# Empty factor returns zero-filled items.
s_summary(factor(levels = c("a", "b", "c")))

## Management of NA values.
x <- factor(c(NA, "Female"))
x <- explicit_na(x)
s_summary(x, na.rm = TRUE)
s_summary(x, na.rm = FALSE)

## Different denominators.
analyze_vars_in_cols

Summarize numeric variables in columns

```r

x <- factor(c("a", "a", "b", "c", "a"))
s_summary(x, denom = "N_row", .N_row = 10L)
s_summary(x, denom = "N_col", .N_col = 20L)

# `s_summary.character`

## Basic usage:
s_summary(c("a", "a", "b", "c", "a"), .var = "x", verbose = FALSE)
s_summary(c("a", "a", "b", "c", "a"), .var = "x", na.rm = FALSE, verbose = FALSE)

# `s_summary.logical`

## Basic usage:
s_summary(c(TRUE, FALSE, TRUE, TRUE))

# Empty factor returns zero-filled items.
s_summary(as.logical(c()))

## Management of NA values.
x <- c(NA, TRUE, FALSE)
s_summary(x, na.rm = TRUE)
s_summary(x, na.rm = FALSE)

## Different denominators.
x <- c(TRUE, FALSE, TRUE, TRUE)
s_summary(x, denom = "N_row", .N_row = 10L)
s_summary(x, denom = "N_col", .N_col = 20L)

a_summary(factor(c("a", "a", "b", "c", "a")), .N_row = 10, .N_col = 10)
a_summary(
  factor(c("a", "a", "b", "c", "a")),
  .ref_group = factor(c("a", "a", "b", "c")), compare = TRUE
)
a_summary(
  c("A", "B", "A", "C"),
  .ref_group = c("B", "A", "C"), .var = "x", compare = TRUE, verbose = FALSE
)
a_summary(c(TRUE, FALSE, FALSE, TRUE, TRUE), .N_row = 10, .N_col = 10)
a_summary(
  c(TRUE, FALSE, FALSE, TRUE, TRUE),
  .ref_group = c(TRUE, FALSE), .in_ref_col = TRUE, compare = TRUE
)
a_summary(rnorm(10), .N_col = 10, .N_row = 20, .var = "bla")
a_summary(rnorm(10, 5, 1), .ref_group = rnorm(20, -5, 1), .var = "bla", compare = TRUE)
```

Description

[Experimental]

Layout-creating function which can be used for creating column-wise summary tables. This function sets the analysis methods as column labels and is a wrapper for \texttt{rtables::analyze_colvars()}. It was designed principally for PK tables.

Usage

\begin{verbatim}
analyze_vars_in_cols(
    lyt,
    vars,
    ...,  
    .stats = c("n", "mean", "sd", "se", "cv", "geom_cv"),
    .labels = c(n = "n", mean = "Mean", sd = "SD", se = "SE", cv = "CV (%)", geom_cv = "CV % Geometric Mean"),
    row_labels = NULL,
    do_summarize_row_groups = FALSE,
    split_col_vars = TRUE,
    imp_rule = NULL,
    avalcat_var = "AVALCAT1",
    cache = FALSE,
    .indent_mods = NULL,
    na_str = default_na_str(),
    nested = TRUE,
    .formats = NULL,
    .aligns = NULL)
\end{verbatim}

Arguments

\begin{itemize}
  \item \texttt{lyt} \hspace{1cm} (PreDataTableLayouts)
    layout that analyses will be added to.
  \item \texttt{vars} \hspace{1cm} (character)
    variable names for the primary analysis variable to be iterated over.
  \item \texttt{...} \hspace{1cm} additional arguments for the lower level functions.
  \item \texttt{.stats} \hspace{1cm} (character)
    statistics to select for the table.
  \item \texttt{.labels} \hspace{1cm} (named character)
    labels for the statistics (without indent).
  \item \texttt{row_labels} \hspace{1cm} (character)
    as this function works in columns space, usually \texttt{.labels} character vector applies on the column space. You can change the row labels by defining this parameter to a named character vector with names corresponding to the split values. It defaults to NULL and if it contains only one string, it will duplicate that as a row label.
\end{itemize}
analyze_vars_in_cols

```r
do_summarize_row_groups
  (flag)
  defaults to FALSE and applies the analysis to the current label rows. This is a
  wrapper of `rtables::summarize_row_groups()` and it can accept labelstr
to define row labels. This behavior is not supported as we never need to overload
row labels.

split_col_vars (flag)
  defaults to TRUE and puts the analysis results onto the columns. This option
  allows you to add multiple instances of this functions, also in a nested fashion,
  without adding more splits. This split must happen only one time on a single
layout.

imp_rule (string or NULL)
  imputation rule setting. Defaults to NULL for no imputation rule. Can also be
  "1/3" to implement 1/3 imputation rule or "1/2" to implement 1/2 imputation
  rule. In order to use an imputation rule, the avalcat_var argument must be
  specified. See `imputation_rule()` for more details on imputation.

avalcat_var (string)
  if imp_rule is not NULL, name of variable that indicates whether a row in the
data corresponds to an analysis value in category "BLQ", "LTR", "<PCLLOQ", or
  none of the above (defaults to "AVALCAT1"). Variable must be present in the data
and should match the variable used to calculate the n_blq statistic (if included
in .stats).

cache (flag)
  whether to store computed values in a temporary caching environment. This
will speed up calculations in large tables, but should be set to FALSE if the same
rtable layout is used for multiple tables with different data. Defaults to FALSE.

.indent_mods (named integer)
  indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified
default behavior. Can be negative.

na_str (string)
  string used to replace all NA or empty values in the output.

nested (flag)
  whether this layout instruction should be applied within the existing layout
structure _if possible (TRUE, the default) or as a new top-level element (FALSE).
Ignored if it would nest a split. underneath analyses, which is not allowed.

.formats (named character or list)
  formats for the statistics. See Details in `analyze_vars` for more information on
the "auto" setting.

.aligns (character or NULL)
  alignment for table contents (not including labels). When NULL, "center" is
  applied. See `formatters::list_valid_aligns()` for a list of all currently
  supported alignments.
```

**Value**

A layout object suitable for passing to further layouting functions, or to `rtables::build_table()`. Adding this function to an `rtable` layout will summarize the given variables, arrange the output in
columns, and add it to the table layout.

**Note**

This is an experimental implementation of `rtables::summarize_row_groups()` and `rtables::analyze_colvars()` that may be subjected to changes as `rtables` extends its support to more complex analysis pipelines on the column space. For the same reasons, we encourage to read the examples carefully and file issues for cases that differ from them.

Here `labelstr` behaves differently than usual. If it is not defined (default as `NULL`), row labels are assigned automatically to the split values in case of `rtables::analyze_colvars` (`do_summarize_row_groups = FALSE`, the default), and to the group label for `do_summarize_row_groups = TRUE`.

**See Also**

`analyze_vars()`, `rtables::analyze_colvars()`.

**Examples**

```r
library(dplyr)

# Data preparation
adpp <- tern_ex_adpp %>% h_pkparam_sort()

lyt <- basic_table() %>%
  split_rows_by(var = "STRATA1", label_pos = "topleft") %>%
  split_rows_by(
    var = "SEX",
    label_pos = "topleft",
    child_labels = "hidden"
  ) %>%
  # Removes duplicated labels
  analyze_vars_in_cols(vars = "AGE")
result <- build_table(lyt = lyt, df = adpp)
result

# By selecting just some statistics and ad-hoc labels
lyt <- basic_table() %>%
  split_rows_by(var = "ARM", label_pos = "topleft") %>%
  split_rows_by(
    var = "SEX",
    label_pos = "topleft",
    child_labels = "hidden",
    split_fun = drop_split_levels
  ) %>%
  analyze_vars_in_cols(
    vars = "AGE",
    .stats = c("n", "cv", "geom_mean"),
    .labels = c(
      n = "aN",
      cv = "aCV",
      geom_mean = "aGeomMean"
    )
  )
```

append_varlabels

result <- build_table(lyt = lyt, df = adpp)
result

# Changing row labels
lyt <- basic_table() %>%
analyze_vars_in_cols(
  vars = "AGE",
  row_labels = "some custom label"
)
result <- build_table(lyt, df = adpp)
result

# Pharmacokinetic parameters
lyt <- basic_table() %>%
split_rows_by(
  var = "TLG_DISPLAY",
  split_label = "PK Parameter",
  label_pos = "topleft",
  child_labels = "hidden"
) %>%
analyze_vars_in_cols(
  vars = "AVAL"
)
result <- build_table(lyt, df = adpp)
result

# Multiple calls (summarize label and analyze underneath)
lyt <- basic_table() %>%
split_rows_by(
  var = "TLG_DISPLAY",
  split_label = "PK Parameter",
  label_pos = "topleft"
) %>%
analyze_vars_in_cols(
  vars = "AVAL",
  do_summarize_row_groups = TRUE # does a summarize level
) %>%
split_rows_by("SEX",
  child_labels = "hidden",
  label_pos = "topleft"
) %>%
analyze_vars_in_cols(
  vars = "AVAL",
  split_col_vars = FALSE # avoids re-splitting the columns
)
result <- build_table(lyt, df = adpp)
result

append_varlabels
Add variable labels to top left corner in table
Description

[Stable]
Helper layout-creating function to append the variable labels of a given variables vector from a
given dataset in the top left corner. If a variable label is not found then the variable name itself is
used instead. Multiple variable labels are concatenated with slashes.

Usage

append_varlabels(lyt, df, vars, indent = 0L)

Arguments

- **lyt** (PreDataTableLayouts)
  layout that analyses will be added to.
- **df** (data.frame)
  data set containing all analysis variables.
- **vars** (character)
  variable names of which the labels are to be looked up in df.
- **indent** (integer(1))
  non-negative number of nested indent space, default to 0L which means no inden-
t. 1L means two spaces indent, 2L means four spaces indent and so on.

Value

A modified layout with the new variable label(s) added to the top-left material.

Note

This is not an optimal implementation of course, since we are using here the data set itself during
the layout creation. When we have a more mature rtables implementation then this will also be
improved or not necessary anymore.

Examples

```r
lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  split_rows_by("SEX") %>%
  append_varlabels(DM, "SEX") %>%
  analyze("AGE", afun = mean) %>%
  append_varlabels(DM, "AGE", indent = 1)
build_table(lyt, DM)

lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  split_rows_by("SEX") %>%
  analyze("AGE", afun = mean) %>%
  append_varlabels(DM, c("SEX", "AGE"))
build_table(lyt, DM)
```
arrange_grobs

Arrange multiple grobs

Description

[Deprecated]

Arrange grobs as a new grob with \( n \times m \) (rows \( \times \) cols) layout.

Usage

```r
arrange_grobs(
    ..., 
    grobs = list(...),
    ncol = NULL,
    nrow = NULL,
    padding_ht = grid::unit(2, "line"),
    padding_wt = grid::unit(2, "line"),
    vp = NULL,
    gp = NULL,
    name = NULL
)
```

Arguments

- `...`: grobs.
- `grobs`: list of grob
  a list of grobs.
- `ncol`: integer(1)
  number of columns in layout.
- `nrow`: integer(1)
  number of rows in layout.
- `padding_ht`: grid::unit
  unit of length 1, vertical space between each grob.
- `padding_wt`: grid::unit
  unit of length 1, horizontal space between each grob.
- `vp`: viewport or NULL
  a viewport() object (or NULL).
- `gp`: gpar
  a gpar() object.
- `name`: string
  a character identifier for the grob.

Value

A grob.
Examples

```r
library(grid)

num <- lapply(1:9, textGrob)
grid::grid.newpage()
grid.draw(arrange_grobs(grobs = num, ncol = 2))
showViewport()

g1 <- circleGrob(gp = gpar(col = "blue"))
g2 <- circleGrob(gp = gpar(col = "red"))
g3 <- textGrob("TEST TEXT")
grid::grid.newpage()
grid.draw(arrange_grobs(g1, g2, g3, nrow = 2))
showViewport()
grid::grid.newpage()
grid.draw(arrange_grobs(g1, g2, g3, ncol = 3))
grid::grid.newpage()
g1 <- grid::viewport(grid::grid.layout(1, 2))
v2 <- grid::viewport(grid::grid.layout(1, 2))
grid.draw(arrange_grobs(g1, g2, ncol = 2, vp = v2))
showViewport()
```

---

**as.rtable**  
*Convert to rtable*

**Description**

[Stable]

This is a new generic function to convert objects to rtable tables.

**Usage**

```r
as.rtable(x, ...
```

### S3 method for class 'data.frame'

```r
as.rtable(x, format = "xx.xx", ...)
```

**Arguments**

- `x`  
  (data.frame)  
  the object which should be converted to an rtable.
... additional arguments for methods.

Value

An rtable table object. Note that the concrete class will depend on the method used.

Methods (by class)

- \texttt{as.rtable(data.frame)}: Method for converting a \texttt{data.frame} that contains numeric columns to rtable.

Examples

```r
x <- data.frame(
a = 1:10,
b = rnorm(10)
)
as.rtable(x)
```

### bland_altman

**Bland-Altman analysis**

**Description**

[Experimental]

Functions that use the Bland-Altman method to assess the agreement between two numerical vectors.

**Usage**

- \texttt{s_bland_altman(x, y, conf_level = 0.95)}
- \texttt{g_bland_altman(x, y, conf_level = 0.95)}

**Arguments**

- \texttt{x} (numeric)
  vector of numbers we want to analyze.
- \texttt{y} (numeric)
  vector of numbers we want to analyze, to be compared with \texttt{x}.
- \texttt{conf_level} (proportion)
  confidence level of the interval.
Value

- `s_bland_altman()` returns a named list of the following elements: `df`, `difference_mean`, `ci_mean`, `difference_sd`, `difference_se`, `upper_agreement_limit`, `lower_agreement_limit`, `agreement_limit_se`, `upper_agreement_limit_ci`, `lower_agreement_limit_ci`, `t_value`, and `n`.

- `g_bland_altman()` returns a ggplot Bland-Altman plot.

Functions

- `s_bland_altman()` : Statistics function that compares two numeric vectors using the Bland-Altman method and calculates a variety of statistics.
- `g_bland_altman()` : Graphing function that produces a Bland-Altman plot.

Examples

```r
x <- seq(1, 60, 5)
y <- seq(5, 50, 4)
conf_level <- 0.9

# Derive statistics that are needed for Bland-Altman plot
s_bland_altman(x, y, conf_level = conf_level)

# Create a Bland-Altman plot
g_bland_altman(x = x, y = y, conf_level = conf_level)
```

combination_function Class for CombinationFunction

Description

[Stable]

CombinationFunction is an S4 class which extends standard functions. These are special functions that can be combined and negated with the logical operators.

Usage

```r
## S4 method for signature 'CombinationFunction,CombinationFunction'
e1 & e2

## S4 method for signature 'CombinationFunction,CombinationFunction'
e1 | e2

## S4 method for signature 'CombinationFunction'
!x
```
**Arguments**

- **e1** (CombinationFunction)
  - left hand side of logical operator.
- **e2** (CombinationFunction)
  - right hand side of logical operator.
- **x** (CombinationFunction)
  - the function which should be negated.

**Value**

A logical value indicating whether the left hand side of the equation equals the right hand side.

**Functions**

- **e1 & e2**: Logical "AND" combination of CombinationFunction functions. The resulting object is of the same class, and evaluates the two argument functions. The result is then the "AND" of the two individual results.
- **e1 | e2**: Logical "OR" combination of CombinationFunction functions. The resulting object is of the same class, and evaluates the two argument functions. The result is then the "OR" of the two individual results.
- `'!'(CombinationFunction)`: Logical negation of CombinationFunction functions. The resulting object is of the same class, and evaluates the original function. The result is then the opposite of this result.

**Examples**

```r
higher <- function(a) {
  force(a)
  CombinationFunction(
    function(x) {
      x > a
    }
  )
}

lower <- function(b) {
  force(b)
  CombinationFunction(
    function(x) {
      x < b
    }
  )
}

c1 <- higher(5)
c2 <- lower(10)
c3 <- higher(5) & lower(10)
c3(7)
```
Combine counts

Description
Simplifies the estimation of column counts, especially when group combination is required.

Usage
combine_counts(fct, groups_list = NULL)

Arguments
- fct (factor): the variable with levels which needs to be grouped.
- groups_list (named list of character): specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.

Value
A vector of column counts.

See Also
combine_groups()

Examples
ref <- c("A: Drug X", "B: Placebo")
groups <- combine_groups(fct = DM$ARM, ref = ref)

col_counts <- combine_counts(fct = DM$ARM, groups_list = groups)

basic_table() %>%
  split_cols_by_groups("ARM", groups) %>%
  add_colcounts() %>%
  analyze_vars("AGE") %>%
  build_table(DM, col_counts = col_counts)

ref <- "A: Drug X"
groups <- combine_groups(fct = DM$ARM, ref = ref)
col_counts <- combine_counts(fct = DM$ARM, groups_list = groups)
**combine_groups**

Reference and treatment group combination

### Description

**[Stable]**

Facilitate the re-combination of groups divided as reference and treatment groups; it helps in arranging groups of columns in the *rtables* framework and teal modules.

### Usage

```r
combine_groups(fct, ref = NULL, collapse = "/")
```

### Arguments

- **fct** *(factor)*
  - the variable with levels which needs to be grouped.
- **ref** *(character)*
  - the reference level(s).
- **collapse** *(string)*
  - a character string to separate *fct* and *ref*.

### Value

A list with first item *ref* (reference) and second item *trt* (treatment).

### Examples

```r
groups <- combine_groups(
  fct = DM$ARM,
  ref = c("B: Placebo")
)

basic_table() %>%
  split_cols_by_groups("ARM", groups) %>%
  add_colcounts() %>%
  analyze_vars("AGE") %>%
  build_table(DM, col_counts = col_counts)
```
**combine_levels** Combine factor levels

**Description**

[Stable]
Combine specified old factor Levels in a single new level.

**Usage**

```
combine_levels(x, levels, new_level = paste(levels, collapse = "/"))
```

**Arguments**

- `x` (factor) factor variable.
- `levels` (character) level names to be combined.
- `new_level` (string) name of new level.

**Value**

A factor with the new levels.

**Examples**

```
x <- factor(letters[1:5], levels = letters[5:1])
combine_levels(x, levels = c("a", "b"))
combine_levels(x, c("e", "b"))
```

**combine_vectors** Element-wise combination of two vectors

**Description**

Element-wise combination of two vectors.

**Usage**

```
combine_vectors(x, y)
```
compare_variables

Arguments

x (vector)
first vector to combine.

y (vector)
second vector to combine.

Value

A list where each element combines corresponding elements of x and y.

Examples

combine_vectors(1:3, 4:6)

---

compare_variables

Compare variables between groups

Description

[Stable]
Comparison with a reference group for different x objects.

Usage

compare_vars(
  lyt,
  vars,
  var_labels = vars,
  na_str = default_na_str(),
  nested = TRUE,
  ...
)

s_compare(x, .ref_group, .in_ref_col, ...)

## S3 method for class 'numeric'
s_compare(x, .ref_group, .in_ref_col, ...)
```r
## S3 method for class 'factor'
s_compare(x, .ref_group, .in_ref_col, denom = "n", na.rm = TRUE, ...)

## S3 method for class 'character'
s_compare(
x,
.ref_group,
.in_ref_col,
denom = "n",
na.rm = TRUE,
.var,
verbose = TRUE,
...
)

## S3 method for class 'logical'
s_compare(x, .ref_group, .in_ref_col, na.rm = TRUE, denom = "n", ...)
```

Arguments

- **lyt** *(PreDataTableLayouts)*
  layout that analyses will be added to.
- **vars** *(character)*
  variable names for the primary analysis variable to be iterated over.
- **var_labels** *(character)*
  variable labels.
- **na_str** *(string)*
  string used to replace all NA or empty values in the output.
- **nested** *(flag)*
  whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.
- **...** arguments passed to `s_compare()`.
- **na.rm** *(flag)*
  whether NA values should be removed from x prior to analysis.
- **show_labels** *(string)*
  label visibility: one of "default", "visible" and "hidden".
- **table_names** *(character)*
  this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.
- **section_div** *(string)*
  string which should be repeated as a section divider after each group defined by this split instruction, or NA_character_ (the default) for no section divider.
- **.stats** *(character)*
  statistics to select for the table. Run `get_stats("analyze_vars_numeric")` to
see statistics available for numeric variables, and `get_stats("analyze_vars_counts")` for statistics available for non-numeric variables.

```
.forms (named character or list)
formats for the statistics. See Details in analyze_vars for more information on
the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Each element of the vector should be a name-
value pair with name corresponding to a statistic specified in .stats and value
the indentation for that statistic's row label.

x (numeric)
vector of numbers we want to analyze.

.ref_group (data.frame or vector)
the data corresponding to the reference group.

.in_ref_col (flag)
TRUE when working with the reference level, FALSE otherwise.

denom (string)
choice of denominator for factor proportions, can only be n (number of values
in this row and column intersection).

.var (string)
single variable name that is passed by rtables when requested by a statistics
function.

.verbose (flag)
whether warnings and messages should be printed. Mainly used to print out
information about factor casting. Defaults to TRUE.
```

**Value**

- `compare_vars()` returns a layout object suitable for passing to further layouting functions, or
to `rtables::build_table()`. Adding this function to an rtable layout will add formatted
rows containing the statistics from `s_compare()` to the table layout.

- `s_compare()` returns output of `s_summary()` and comparisons versus the reference group in
the form of p-values.

**Functions**

- `compare_vars()`: Layout-creating function which can take statistics function arguments and
additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `s_compare()`: S3 generic function to produce a comparison summary.
- `s_compare(numeric)`: Method for numeric class. This uses the standard t-test to calculate
the p-value.
- `s_compare(factor)`: Method for factor class. This uses the chi-squared test to calculate
the p-value.
compare_variables

- **s_compare(character)**: Method for character class. This makes an automatic conversion to factor (with a warning) and then forwards to the method for factors.
- **s_compare(logical)**: Method for logical class. A chi-squared test is used. If missing values are not removed, then they are counted as FALSE.

**Note**

- For factor variables, denom for factor proportions can only be n since the purpose is to compare proportions between columns, therefore a row-based proportion would not make sense. Proportion based on N_col would be difficult since we use counts for the chi-squared test statistic, therefore missing values should be accounted for as explicit factor levels.
- If factor variables contain NA, these NA values are excluded by default. To include NA values set na.rm = FALSE and missing values will be displayed as an NA level. Alternatively, an explicit factor level can be defined for NA values during pre-processing via `df_explicit_na()` - the default na_level ("<Missing>") will also be excluded when na.rm is set to TRUE.
- For character variables, automatic conversion to factor does not guarantee that the table will be generated correctly. In particular for sparse tables this very likely can fail. Therefore it is always better to manually convert character variables to factors during pre-processing.
- For `compare_vars()`, the column split must define a reference group via ref_group so that the comparison is well defined.

**See Also**

- `s_summary()` which is used internally to compute a summary within `s_compare()`, and `a_summary()` which is used (with compare = TRUE) as the analysis function for `compare_vars()`.

**Examples**

```r
# `compare_vars()` in `rtables` pipelines

## Default output within a `rtables` pipeline.
lyt <- basic_table() %>%
  split_cols_by("ARMCD", ref_group = "ARM B") %>%
  compare_vars(c("AGE", "SEX"))
build_table(lyt, tern_ex_adsl)

## Select and format statistics output.
lyt <- basic_table() %>%
  split_cols_by("ARMCD", ref_group = "ARM C") %>%
  compare_vars(
    vars = "AGE",
    .stats = c("mean_sd", "pval"),
    .formats = c(mean_sd = "xx.x, xx.x"),
    .labels = c(mean_sd = "Mean, SD")
  )
build_table(lyt, df = tern_ex_adsl)

# `s_compare.numeric`

## Usual case where both this and the reference group vector have more than 1 value.
```
s_compare(rnorm(10, 5, 1), .ref_group = rnorm(5, -5, 1), .in_ref_col = FALSE)

## If one group has not more than 1 value, then p-value is not calculated.
s_compare(rnorm(10, 5, 1), .ref_group = 1, .in_ref_col = FALSE)

## Empty numeric does not fail, it returns NA-filled items and no p-value.
s_compare(numeric(), .ref_group = numeric(), .in_ref_col = FALSE)

# `s_compare.factor`

## Basic usage:
x <- factor(c("a", "a", "b", "c", "a"))
y <- factor(c("a", "b", "c"))
s_compare(x = x, .ref_group = y, .in_ref_col = FALSE)

## Management of NA values.
x <- explicit_na(factor(c("a", "a", "b", "c", "a", NA, NA)))
y <- explicit_na(factor(c("a", "b", "c", NA)))
s_compare(x = x, .ref_group = y, .in_ref_col = FALSE, na.rm = TRUE)
s_compare(x = x, .ref_group = y, .in_ref_col = FALSE, na.rm = FALSE)

# `s_compare.character`

## Basic usage:
x <- c("a", "a", "b", "c", "a")
y <- c("a", "b", "c")
s_compare(x, .ref_group = y, .in_ref_col = FALSE, .var = "x", verbose = FALSE)

## Note that missing values handling can make a large difference:
x <- c("a", "a", "b", "c", "a", NA)
y <- c("a", "b", "c", rep(NA, 20))
s_compare(x,
  .ref_group = y, .in_ref_col = FALSE,
  .var = "x", verbose = FALSE)
)  
s_compare(x,
  .ref_group = y, .in_ref_col = FALSE, .var = "x",
  na.rm = FALSE, verbose = FALSE)
)

# `s_compare.logical`

## Basic usage:
x <- c(TRUE, FALSE, TRUE, TRUE)
y <- c(FALSE, FALSE, TRUE)
s_compare(x, .ref_group = y, .in_ref_col = FALSE)

## Management of NA values.
x <- c(NA, TRUE, FALSE)
y <- c(NA, NA, NA, NA, FALSE)
s_compare(x, .ref_group = y, .in_ref_col = FALSE, na.rm = TRUE)
s_compare(x, .ref_group = y, .in_ref_col = FALSE, na.rm = FALSE)
Description

[Stable]
Sets a list of parameters for summaries of descriptive statistics. Typically used internally to specify details for `s_summary()`. This function family is mainly used by `analyze_vars()`.

Usage

```r
control_analyze_vars(
  conf_level = 0.95,
  quantiles = c(0.25, 0.75),
  quantile_type = 2,
  test_mean = 0
)
```

Arguments

- `conf_level` (proportion)
  confidence level of the interval.
- `quantiles` (numeric(2))
  vector of length two to specify the quantiles to calculate.
- `quantile_type` (numeric(1))
  number between 1 and 9 selecting quantile algorithms to be used. Default is set to 2 as this matches the default quantile algorithm in SAS proc univariate set by QNTLDEF=5. This differs from R’s default. See more about type in `stats::quantile()`.
- `test_mean` (numeric(1))
  number to test against the mean under the null hypothesis when calculating p-value.

Value

A list of components with the same names as the arguments.
Control functions for Kaplan-Meier plot annotation tables

Description

[Stable]
Auxiliary functions for controlling arguments for formatting the annotation tables that can be added to plots generated via `g_km()`.

Usage

```r
control_surv_med_annot(x = 0.8, y = 0.85, w = 0.32, h = 0.16, fill = TRUE)
control_coxph_annot(
  x = 0.29,
  y = 0.51,
  w = 0.4,
  h = 0.125,
  fill = TRUE,
  ref_lbls = FALSE
)
```

Arguments

- `x` (proportion): x-coordinate for center of annotation table.
- `y` (proportion): y-coordinate for center of annotation table.
- `w` (proportion): relative width of the annotation table.
- `h` (proportion): relative height of the annotation table.
- `fill` (flag or character): whether the annotation table should have a background fill color. Can also be a color code to use as the background fill color. If TRUE, color code defaults to "#00000020".
- `ref_lbls` (flag): whether the reference group should be explicitly printed in labels for the annotation table. If FALSE (default), only comparison groups will be printed in the table labels.

Value

A list of components with the same names as the arguments.
Functions

- `control_surv_med_annot()`: Control function for formatting the median survival time annotation table. This annotation table can be added in `g_km()` by setting `annot_surv_med=TRUE`, and can be configured using the `control_surv_med_annot()` function by setting it as the `control_annot_surv_med` argument.

- `control_coxph_annot()`: Control function for formatting the Cox-PH annotation table. This annotation table can be added in `g_km()` by setting `annot_coxph=TRUE`, and can be configured using the `control_coxph_annot()` function by setting it as the `control_annot_coxph` argument.

See Also

`g_km()`

Examples

```r
control_surv_med_annot()
control_coxph_annot()
```

---

`control_coxph`  
*Control function for Cox-PH model*

Description

[Stable]

This is an auxiliary function for controlling arguments for Cox-PH model, typically used internally to specify details of Cox-PH model for `s_coxph_pairwise()`. `conf_level` refers to Hazard Ratio estimation.

Usage

```r
control_coxph(
  pval_method = c("log-rank", "wald", "likelihood"),
  ties = c("efron", "breslow", "exact"),
  conf_level = 0.95
)
```

Arguments

- `pval_method` *(string)*
  p-value method for testing hazard ratio = 1. Default method is "log-rank", can also be set to "wald" or "likelihood".

- `ties` *(string)*
  string specifying the method for tie handling. Default is "efron", can also be set to "breslow" or "exact". See more in `survival::coxph()`.
conf_level (proportion)
  confidence level of the interval.

Value

A list of components with the same names as the arguments.

control_coxreg

Control function for Cox regression

Description

[Stable]

Sets a list of parameters for Cox regression fit. Used internally.

Usage

control_coxreg(
  pval_method = c("wald", "likelihood"),
  ties = c("exact", "efron", "breslow"),
  conf_level = 0.95,
  interaction = FALSE
)

Arguments

pval_method (string)
  the method used for estimation of p.values; wald (default) or likelihood.

ties (string)
  among exact (equivalent to DISCRETE in SAS), efron and breslow, see survival::coxph().
  Note: there is no equivalent of SAS EXACT method in R.

conf_level (proportion)
  confidence level of the interval.

interaction (flag)
  if TRUE, the model includes the interaction between the studied treatment and candidate covariate. Note that for univariate models without treatment arm, and multivariate models, no interaction can be used so that this needs to be FALSE.

Value

A list of items with names corresponding to the arguments.

See Also

fit_coxreg_univar() and fit_coxreg_multivar().
control_incidence_rate

Control function for incidence rate

Description

[Stable]
This is an auxiliary function for controlling arguments for the incidence rate, used internally to specify details in s_incidence_rate().

Usage

```r
control_incidence_rate(
  conf_level = 0.95,
  conf_type = c("normal", "normal_log", "exact", "byar"),
  input_time_unit = c("year", "day", "week", "month"),
  num_pt_year = 100
)
```

Arguments

- `conf_level` (proportion)
  confidence level of the interval.
- `conf_type` (string)
  normal (default), normal_log, exact, or byar for confidence interval type.
- `input_time_unit` (string)
  day, week, month, or year (default) indicating time unit for data input.
- `num_pt_year` (numeric(1))
  number of patient-years to use when calculating adverse event rates.

Value

A list of components with the same names as the arguments.

See Also

incidence_rate

Examples

```r
control_incidence_rate(0.9, "exact", "month", 100)
```
Description

[Stable]
Default values for variables parameter in g_lineplot function. A variable’s default value can be overwritten for any variable.

Usage

control_lineplot_vars(
x = "AVISIT",
y = "AVAL",
group_var = "ARM",
facet_var = NA,
paramcd = "PARAMCD",
y_unit = "AVALU",
subject_var = "USUBJID",
strata = lifecycle::deprecated(),
cohort_id = lifecycle::deprecated()
)

Arguments

x (string)
x-variable name.

y (string)
y-variable name.

group_var (string or NA)
group variable name.

facet_var (string or NA)
faceting variable name.

paramcd (string or NA)
parameter code variable name.

y_unit (string or NA)
y-axis unit variable name.

subject_var (string or NA)
subject variable name.

strata [Deprecated] use the group_var parameter instead.

cohort_id [Deprecated] use the subject_var parameter instead.

Value

A named character vector of variable names.
control_logistic

Control function for logistic regression model fitting

Description

[Stable]
This is an auxiliary function for controlling arguments for logistic regression models. conf_level refers to the confidence level used for the Odds Ratio CIs.

Usage

control_logistic(response_definition = "response", conf_level = 0.95)

Arguments

  response_definition
    (string)
    the definition of what an event is in terms of response. This will be used when fitting the logistic regression model on the left hand side of the formula. Note that the evaluated expression should result in either a logical vector or a factor with 2 levels. By default this is just "response" such that the original response variable is used and not modified further.

  conf_level
    (proportion)
    confidence level of the interval.

Value

A list of components with the same names as the arguments.

Examples

  # Standard options.
  control_logistic()

  # Modify confidence level.
  control_logistic(conf_level = 0.9)

  # Use a different response definition.
  control_logistic(response_definition = "I(response %in% c('CR', 'PR'))")
## control_step

**Control function for subgroup treatment effect pattern (STEP) calculations**

### Description

[Stable]

This is an auxiliary function for controlling arguments for STEP calculations.

### Usage

```r
control_step(
  biomarker = NULL,
  use_percentile = TRUE,
  bandwidth,
  degree = 0L,
  num_points = 39L
)
```

### Arguments

- **biomarker** (numeric or NULL)
  - optional provision of the numeric biomarker variable, which could be used to infer bandwidth, see below.

- **use_percentile** (flag)
  - if TRUE, the running windows are created according to quantiles rather than actual values, i.e. the bandwidth refers to the percentage of data covered in each window. Suggest TRUE if the biomarker variable is not uniformly distributed.

- **bandwidth** (numeric(1) or NULL)
  - indicating the bandwidth of each window. Depending on the argument `use_percentile`, it can be either the length of actual-value windows on the real biomarker scale, or percentage windows. If `use_percentile = TRUE`, it should be a number between 0 and 1. If NULL, treat the bandwidth to be infinity, which means only one global model will be fitted. By default, 0.25 is used for percentage windows and one quarter of the range of the biomarker variable for actual-value windows.

- **degree** (integer(1))
  - the degree of polynomial function of the biomarker as an interaction term with the treatment arm fitted at each window. If 0 (default), then the biomarker variable is not included in the model fitted in each biomarker window.

- **num_points** (integer(1))
  - the number of points at which the hazard ratios are estimated. The smallest number is 2.

### Value

A list of components with the same names as the arguments, except `biomarker` which is just used to calculate the `bandwidth` in case that actual biomarker windows are requested.
control_surv_time

Control function for survfit models for survival time

Examples

# Provide biomarker values and request actual values to be used, # so that bandwidth is chosen from range.
control_step(biomarker = 1:10, use_percentile = FALSE)

# Use a global model with quadratic biomarker interaction term.
control_step(bandwidth = NULL, degree = 2)

# Reduce number of points to be used.
control_step(num_points = 10)

control_surv_time

Description

[Stable]

This is an auxiliary function for controlling arguments for survfit model, typically used internally to specify details of survfit model for s_surv_time(). conf_level refers to survival time estimation.

Usage

control_surv_time(
  conf_level = 0.95,
  conf_type = c("plain", "log", "log-log"),
  quantiles = c(0.25, 0.75)
)

Arguments

conf_level (proportion)
  confidence level of the interval.

conf_type (string)
  confidence interval type. Options are "plain" (default), "log", "log-log", see more in survival::survfit(). Note option "none" is no longer supported.

quantiles (numeric(2))
  vector of length two specifying the quantiles of survival time.

Value

A list of components with the same names as the arguments.
control_surv_timepoint

Control function for survfit models for patients’ survival rate at time points

Description

[Stable]

This is an auxiliary function for controlling arguments for survfit model, typically used internally to specify details of survfit model for `s_surv_timepoint()`. conf_level refers to patient risk estimation at a time point.

Usage

```r
control_surv_timepoint(
  conf_level = 0.95,
  conf_type = c("plain", "log", "log-log")
)
```

Arguments

- `conf_level` (proportion)
  confidence level of the interval.
- `conf_type` (string)
  confidence interval type. Options are "plain" (default), "log", "log-log", see more in `survival::survfit()`. Note option "none" is no longer supported.

Value

A list of components with the same names as the arguments.

count_occurrences  Occurrence counts

Description

[Stable]

Functions for analyzing frequencies and fractions of occurrences for patients with occurrence data. Primary analysis variables are the dictionary terms. All occurrences are counted for total counts. Multiple occurrences within patient at the lowest term level displayed in the table are counted only once.
Usage

count_occurrences(
    lyt,
    vars,
    id = "USUBJID",
    drop = TRUE,
    var_labels = vars,
    show_labels = "hidden",
    riskdiff = FALSE,
    na_str = default_na_str(),
    nested = TRUE,
    ...
    table_names = vars,
    .stats = "count_fraction_fixed_dp",
    .formats = NULL,
    .labels = NULL,
    .indent_mods = NULL
)

summarize_occurrences(
    lyt,
    var,
    id = "USUBJID",
    drop = TRUE,
    riskdiff = FALSE,
    na_str = default_na_str(),
    ...
    .stats = "count_fraction_fixed_dp",
    .formats = NULL,
    .indent_mods = NULL,
    .labels = NULL
)

s_count_occurrences(
    df,
    denom = c("N_col", "n"),
    .N_col,
    .df_row,
    drop = TRUE,
    .var = "MHDECOD",
    id = "USUBJID"
)

a_count_occurrences(
    df,
    labelstr = "",
    id = "USUBJID",
    denom = c("N_col", "n"),
drop = TRUE,
.N_col,
.var = NULL,
.df_row = NULL,
.stats = NULL,
.formats = NULL,
.labels = NULL,
.indent_mods = NULL,
na_str = default_na_str()
)

Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

vars (character)
variable names for the primary analysis variable to be iterated over.

id (string)
subject variable name.

drop (flag)
whether non-appearing occurrence levels should be dropped from the resulting table. Note that in that case the remaining occurrence levels in the table are sorted alphabetically.

var_labels (character)
variable labels.

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

riskdiff (flag)
whether a risk difference column is present. When set to TRUE, add_riskdiff() must be used as split_fun in the prior column split of the table layout, specifying which columns should be compared. See stat_propdiff_ci() for details on risk difference calculation.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

table_names (character)
this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("count_occurrences") to see available statistics for this function.
count_occurrences

.forms (named character or list)
formats for the statistics. See Details in analyze_vars for more information on
the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodi-
ﬁed default behavior. Can be negative.

df (data.frame)
data set containing all analysis variables.

denom (string)
choice of denominator for patient proportions. Can be:
  • N_col: total number of patients in this column across rows
  • n: number of patients with any occurrences

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typi-
cally passed by rtables.

.df_row (data.frame)
data frame across all of the columns for the given row split.

.var, var (string)
single variable name that is passed by rtables when requested by a statistics
function.

.labelstr (string)
label of the level of the parent split currently being summarized (must be present
as second argument in Content Row Functions). See rtables::summarize_row_groups() for more information.

Value

• count_occurrences() returns a layout object suitable for passing to further layouting func-
tions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_count_occurrences() to the table layout.

• summarize_occurrences() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted content rows containing the statistics from s_count_occurrences() to the table layout.

• s_count_occurrences() returns a list with:
  – count: list of counts with one element per occurrence.
  – count_fraction: list of counts and fractions with one element per occurrence.
  – fraction: list of numerators and denominators with one element per occurrence.

• a_count_occurrences() returns the corresponding list with formatted rtables::CellValue().
Functions

- `count_occurrences()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `summarize_occurrences()`: Layout-creating function which can take content function arguments and additional format arguments. This function is a wrapper for `rtables::summarize_row_groups()`.
- `s_count_occurrences()`: Statistics function which counts number of patients that report an occurrence.
- `a_count_occurrences()`: Formatted analysis function which is used as `afun` in `count_occurrences()`.

Note

By default, occurrences which don’t appear in a given row split are dropped from the table and the occurrences in the table are sorted alphabetically per row split. Therefore, the corresponding layout needs to use `split_fun = drop_split_levels` in the `split_rows_by` calls. Use `drop = FALSE` if you would like to show all occurrences.

Examples

```r
library(dplyr)

df <- data.frame(
  USUBJID = as.character(c(1, 1, 2, 4, 4, 4, 6, 6, 6, 7, 7, 8)),
  ARM = rep(c("A", "B"), each = 6),
)

df_adsl <- df %>%
  select(USUBJID, ARM) %>%
  unique()

# Create table layout
lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  count_occurrences(vars = "MHDECOD", .stats = c("count_fraction"))

# Apply table layout to data and produce `rtable` object
tbl <- lyt %>%
  build_table(df, alt_counts_df = df_adsl) %>%
  prune_table()

tbl

# Layout creating function with custom format.
basic_table() %>%
```
add_colcounts() %>%
split_rows_by("SEX", child_labels = "visible") %>%
summarize_occurrences(
  var = "MHDECOD",
  .formats = c("count_fraction" = "xx.xx (xx.xx%)")
) %>%
build_table(df, alt_counts_df = df_adsl)

# Count unique occurrences per subject.
s_count_occurrences(
  df,
  .N_col = 4L,
  .df_row = df,
  .var = "MHDECOD",
  id = "USUBJID"
)

a_count_occurrences(
  df,
  .N_col = 4L,
  .df_row = df,
  .var = "MHDECOD",
  id = "USUBJID"
)

---

count_occurrences_by_grade

Occurrence counts by grade

Description

[Stable]

Functions for analyzing frequencies and fractions of occurrences by grade for patients with occurrence data. Multiple occurrences within one individual are counted once at the greatest intensity/highest grade level.

Usage

count_occurrences_by_grade(
  lyt,
  var,
  id = "USUBJID",
  grade_groups = list(),
  remove_single = TRUE,
  only_grade_groups = FALSE,
  var_labels = var,
  show_labels = "default",
  riskdiff = FALSE,
count_occurrences_by_grade

    na_str = default_na_str(),
    nested = TRUE,
    ...,
    table_names = var,
    .stats = NULL,
    .formats = NULL,
    .indent_mods = NULL,
    .labels = NULL
)

summarize_occurrences_by_grade(
    lyt,
    var,
    id = "USUBJID",
    grade_groups = list(),
    remove_single = TRUE,
    only_grade_groups = FALSE,
    na_str = default_na_str(),
    ...,
    .stats = NULL,
    .formats = NULL,
    .indent_mods = NULL,
    .labels = NULL
)

s_count_occurrences_by_grade(
    df,
    .var,
    .N_col,
    id = "USUBJID",
    grade_groups = list(),
    remove_single = TRUE,
    only_grade_groups = FALSE,
    labelstr = ""
)

a_count_occurrences_by_grade(
    df,
    .var,
    .N_col,
    id = "USUBJID",
    grade_groups = list(),
    remove_single = TRUE,
    only_grade_groups = FALSE,
    labelstr = ""
count_occurrences_by_grade

Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

id (string)
subject variable name.

grade_groups (named list of character)
list containing groupings of grades.

remove_single (flag)
TRUE to not include the elements of one-element grade groups in the output list; in this case only the grade groups names will be included in the output. If only_grade_groups is set to TRUE this argument is ignored.

only_grade_groups (flag)
whether only the specified grade groups should be included, with individual grade rows removed (TRUE), or all grades and grade groups should be displayed (FALSE).

var_labels (character)
variable labels.

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

riskdiff (flag)
whether a risk difference column is present. When set to TRUE, add_riskdiff() must be used as split_fun in the prior column split of the table layout, specifying which columns should be compared. See stat_propdiff_ci() for details on risk difference calculation.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

table_names (character)
this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("count_occurrences_by_grade") to see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.
count_occurrences_by_grade

.labels (named character)
labels for the statistics (without indent).

df (data.frame)
data set containing all analysis variables.

.var, var (string)
single variable name that is passed by rtables when requested by a statistics function.

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typically passed by rtables.

labelstr (string)
label of the level of the parent split currently being summarized (must be present as second argument in Content Row Functions). See rtables::summarize_row_groups() for more information.

Value

• count_occurrences_by_grade() returns a layout object suitable for passing to further lay-
outing functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_count_occurrences_by_grade() to the table layout.

• summarize_occurrences_by_grade() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted content rows containing the statistics from s_count_occurrences_by_grade() to the table layout.

• s_count_occurrences_by_grade() returns a list of counts and fractions with one element per grade level or grade level grouping.

• a_count_occurrences_by_grade() returns the corresponding list with formatted rtables::CellValue().

Functions

• count_occurrences_by_grade(): Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for rtables::analyze().

• summarize_occurrences_by_grade(): Layout-creating function which can take content function arguments and additional format arguments. This function is a wrapper for rtables::summarize_row_groups().

• s_count_occurrences_by_grade(): Statistics function which counts the number of patients by highest grade.

• a_count_occurrences_by_grade(): Formatted analysis function which is used as afun in count_occurrences_by_grade().

See Also

Relevant helper function h_append_grade_groups().
Examples

library(dplyr)

df <- data.frame(
  USUBJID = as.character(c(1:6, 1)),
  ARM = factor(c("A", "A", "A", "B", "B", "A"), levels = c("A", "B")),
  AETOXGR = factor(c(1, 2, 3, 4, 1, 2, 3), levels = c(1:5)),
  AESEV = factor(
    x = c("MILD", "MODERATE", "SEVERE", "MILD", "MILD", "MODERATE", "SEVERE"),
    levels = c("MILD", "MODERATE", "SEVERE"))
),
stringsAsFactors = FALSE)

df_adsl <- df %>%
  select(USUBJID, ARM) %>%
  unique()

# Layout creating function with custom format.
basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  count_occurrences_by_grade(
    var = "AESEV",
    .formats = c("count_fraction" = "xx.xx (xx.xx%)")
  ) %>%
  build_table(df, alt_counts_df = df_adsl)

# Define additional grade groupings.
grade_groups <- list(
  "-Any-" = c("1", "2", "3", "4", "5"),
  "Grade 1-2" = c("1", "2"),
  "Grade 3-5" = c("3", "4", "5")
)

basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  count_occurrences_by_grade(
    var = "AETOXGR",
    grade_groups = grade_groups,
    only_grade_groups = TRUE
  ) %>%
  build_table(df, alt_counts_df = df_adsl)

# Layout creating function with custom format.
basic_table() %>%
  add_colcounts() %>%
  split_rows_by("ARM", child_labels = "visible", nested = TRUE) %>%
  summarize_occurrences_by_grade(
    var = "AESEV",
    .formats = c("count_fraction" = "xx.xx (xx.xx%)")
count_patients_with_event

Count the number of patients with a particular event

Description
[Stable]
The primary analysis variable .var denotes the unique patient identifier.

Usage

count_patients_with_event(
  lyt,
  vars,
  filters,
  riskdiff = FALSE,
  na_str = default_na_str(),
)
nested = TRUE,

...,
table_names = vars,
.stats = "count_fraction",
.formats = NULL,
.labels = NULL,
.indent_mods = NULL
)

s_count_patients_with_event(
  df,
  .var,
  filters,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)

a_count_patients_with_event(
  df,
  .var,
  filters,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)

Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

vars (character)
variable names for the primary analysis variable to be iterated over.

filters (character)
a character vector specifying the column names and flag variables to be used for counting the number of unique identifiers satisfying such conditions. Multiple column names and flags are accepted in this format c("column_name1" = "flag1", "column_name2" = "flag2"). Note that only equality is being accepted as condition.

riskdiff (flag)
whether a risk difference column is present. When set to TRUE, add_riskdiff() must be used as split_fun in the prior column split of the table layout, specifying which columns should be compared. See stat_propdiff_ci() for details on risk difference calculation.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout
structure (if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. Underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

table_names (character)
this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("count_patients_with_event") to see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent.mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

df (data.frame)
data set containing all analysis variables.

.var (string)
name of the column that contains the unique identifier.

.N.col (integer(1))
column-wise N (column count) for the full column being analyzed that is typically passed by rtables.

.N.row (integer(1))
row-wise N (row group count) for the group of observations being analyzed (i.e. with no column-based subsetting) that is typically passed by rtables.

denom (string)
choice of denominator for proportion. Options are:

- n: number of values in this row and column intersection.
- N.row: total number of values in this row across columns.
- N.col: total number of values in this column across rows.

Value

- count_patients_with_event() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_count_patients_with_event() to the table layout.

- s_count_patients_with_event() returns the count and fraction of unique identifiers with the defined event.

- a_count_patients_with_event() returns the corresponding list with formatted rtables::CellValue().
Functions

- `count_patients_with_event()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `s_count_patients_with_event()`: Statistics function which counts the number of patients for which the defined event has occurred.
- `a_count_patients_with_event()`: Formatted analysis function which is used as `afun` in `count_patients_with_event()`.

See Also

`count_patients_with_flags`

Examples

```r
# `count_patients_with_event()`
lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  count_values(
    "STUDYID",
    values = "AB12345",
    .stats = "count",
    .labels = c(count = "Total AEs")
  ) %>%
  count_patients_with_event(
    "SUBJID",
    filters = c("TRTEMFL" = "Y"),
    .labels = c(count_fraction = "Total number of patients with at least one adverse event"),
    table_names = "tbl_all"
  ) %>%
  count_patients_with_event(
    "SUBJID",
    filters = c("TRTEMFL" = "Y", "AEOUT" = "FATAL"),
    .labels = c(count_fraction = "Total number of patients with fatal AEs"),
    table_names = "tbl_fatal"
  ) %>%
  count_patients_with_event(
    "SUBJID",
    filters = c("TRTEMFL" = "Y", "AEOUT" = "FATAL", "AEREL" = "Y"),
    .labels = c(count_fraction = "Total number of patients with related fatal AEs"),
    .indent_mods = c(count_fraction = 2L),
    table_names = "tbl_rel_fatal"
  )

build_table(lyt, tern_ex_adae, alt_counts_df = tern_ex_adsl)

# `s_count_patients_with_event()`

s_count_patients_with_event(
  tern_ex_adae,
  tern_ex_adae,
)
count_patients_with_flags

Count the number of patients with particular flags

Description

[Stable]

The primary analysis variable .var denotes the unique patient identifier.

Usage

count_patients_with_flags(
  lyt,
  var,
  flag_variables,
  flag_labels = NULL,
  var_labels = var,
  show_labels = "hidden",
  riskdiff = FALSE,
  na_str = default_na_str(),
)
count_patients_with_flags

nested = TRUE,
...,
table_names = paste0("tbl_flags_", var),
.stats = "count_fraction",
.formats = NULL,
.indent_mods = NULL
)

s_count_patients_with_flags(
  df,
  .var,
  flag_variables,
  flag_labels = NULL,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)

a_count_patients_with_flags(
  df,
  .var,
  flag_variables,
  flag_labels = NULL,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)

Arguments

lyt (PreDataTableLayouts)
  layout that analyses will be added to.

var (string)
  single variable name that is passed by rtables when requested by a statistics function.

flag_variables (character)
  a vector specifying the names of logical variables from analysis dataset used for counting the number of unique identifiers.

flag_labels (character)
  vector of labels to use for flag variables.

var_labels (character)
  variable labels.

show_labels (string)
  label visibility: one of "default", "visible" and "hidden".

riskdiff (flag)
  whether a risk difference column is present. When set to TRUE, add_riskdiff()
count_patients_with_flags

must be used as split_fun in the prior column split of the table layout, specifying which columns should be compared. See stat_propdiff_ci() for details on risk difference calculation.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout structure if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

table_names (character)
this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("count_patients_with_flags") to see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

df (data.frame)
data set containing all analysis variables.

.var (string)
name of the column that contains the unique identifier.

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typically passed by rtables.

.N_row (integer(1))
row-wise N (row group count) for the group of observations being analyzed (i.e. with no column-based subsetting) that is typically passed by rtables.

denom (string)
choice of denominator for proportion. Options are:

* n: number of values in this row and column intersection.
* N_row: total number of values in this row across columns.
* N_col: total number of values in this column across rows.

Value

* count_patients_with_flags() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_count_patients_with_flags() to the table layout.
count_patients_with_flags

• `s_count_patients_with_flags()` returns the count and the fraction of unique identifiers with each particular flag as a list of statistics `n`, `count`, `count_fraction`, and `n_blq`, with one element per flag.

• `a_count_patients_with_flags()` returns the corresponding list with formatted `rtables::CellValue()`.

Functions

• `count_patients_with_flags()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.

• `s_count_patients_with_flags()`: Statistics function which counts the number of patients for which a particular flag variable is TRUE.

• `a_count_patients_with_flags()`: Formatted analysis function which is used as `afun` in `count_patients_with_flags()`.

Note

If `flag_labels` is not specified, variables labels will be extracted from `df`. If variables are not labeled, variable names will be used instead. Alternatively, a named vector can be supplied to `flag_variables` such that within each name-value pair the name corresponds to the variable name and the value is the label to use for this variable.

See Also

`count_patients_with_event`

Examples

```r
library(dplyr)

# Add labelled flag variables to analysis dataset.
adae <- tern_ex_adae %>%
  mutate(
    fl1 = TRUE %>% with_label("Total AEs"),
    fl2 = (TRTEMFL == "Y") %>%
      with_label("Total number of patients with at least one adverse event"),
    fl3 = (TRTEMFL == "Y" & AEOUT == "FATAL") %>%
      with_label("Total number of patients with fatal AEs"),
    fl4 = (TRTEMFL == "Y" & AEOUT == "FATAL" & AEREL == "Y") %>%
      with_label("Total number of patients with related fatal AEs")
  )

# `count_patients_with_flags`

lyt2 <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  count_patients_with_flags(
    "SUBJID",
    flag_variables = c("fl1", "fl2", "fl3", "fl4"),
    denom = "N_col"
  )
```

build_table(lyt2, adae, alt_counts_df = tern_ex_ads)

# `s_count_patients_with_flags()`

s_count_patients_with_flags(
  adae,
  "SUBJID",
  flag_variables = c("fl1", "fl2", "fl3", "fl4"),
  denom = "N.col",
  .N.col = 1000
)

# We need to ungroup `count_fraction` first so that the `rtables` formatting
# function `format_count_fraction()` can be applied correctly.

# `a_count_patients_with_flags()`

afun <- make_afun(a_count_patients_with_flags,
  .stats = "count_fraction",
  .ungroup_stats = "count_fraction"
)

afun(
  adae,
  .N.col = 10L,  
  .N.row = 10L,  
  .var = "USUBJID",
  flag_variables = c("fl1", "fl2", "fl3", "fl4")
)

---

**count_values_funs**  
*Count specific values*

**Description**

[Stable]

We can count the occurrence of specific values in a variable of interest.

**Usage**

```r
count_values(
  lyt,  
  vars,  
  values,  
  na_str = default_na_str(),  
  nested = TRUE,  
  ...,  
)```
```r
table_names = vars,
.stats = "count_fraction",
.formats = NULL,
.labels = c(count_fraction = paste(values, collapse = ", ")),
.indent_mods = NULL
s_count_values(
  x,
  values,
  na.rm = TRUE,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)
```

## S3 method for class 'character'
```r
s_count_values(x, values = "Y", na.rm = TRUE, ...)
```

## S3 method for class 'factor'
```r
s_count_values(x, values = "Y", ...)
```

## S3 method for class 'logical'
```r
s_count_values(x, values = TRUE, ...)
```

```r
a_count_values(
  x,
  values,
  na.rm = TRUE,
  .N_col,
  .N_row,
  denom = c("n", "N_row", "N_col")
)
```

### Arguments

- **lyt** *(PreDataTableLayouts)*
  - layout that analyses will be added to.
- **vars** *(character)*
  - variable names for the primary analysis variable to be iterated over.
- **values** *(character)*
  - specific values that should be counted.
- **na_str** *(string)*
  - string used to replace all NA or empty values in the output.
- **nested** *(flag)*
  - whether this layout instruction should be applied within the existing layout structure if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.
... additional arguments for the lower level functions.

table_names (character)
this can be customized in the case that the same vars are analyzed multiple
times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("count_values") to see avail-
able statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on
the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodi-
fied default behavior. Can be negative.

x (numeric)
vector of numbers we want to analyze.

na.rm (flag)
whether NA values should be removed from x prior to analysis.

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typi-
cally passed by rtables.

.N_row (integer(1))
row-wise N (row group count) for the group of observations being analyzed (i.e.
with no column-based subsetting) that is typically passed by rtables.

denom (string)
choice of denominator for proportion. Options are:

- n: number of values in this row and column intersection.
- N_row: total number of values in this row across columns.
- N_col: total number of values in this column across rows.

Value

- count_values() returns a layout object suitable for passing to further laying functions, or
to rtables::build_table(). Adding this function to an rtable layout will add formatted
rows containing the statistics from s_count_values() to the table layout.

- s_count_values() returns output of s_summary() for specified values of a non-numeric vari-
able.

- a_count_values() returns the corresponding list with formatted rtables::CellValue().
Functions

- `count_values()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `s_count_values()`: S3 generic function to count values.
- `s_count_values(character)`: Method for character class.
- `s_count_values(factor)`: Method for factor class. This makes an automatic conversion to character and then forwards to the method for characters.
- `s_count_values(logical)`: Method for logical class.
- `a_count_values()`: Formatted analysis function which is used as `afun` in `count_values()`.

Note

- For factor variables, `s_count_values` checks whether values are all included in the levels of `x` and fails otherwise.
- For `count_values()`, variable labels are shown when there is more than one element in `vars`, otherwise they are hidden.

Examples

```r
# `count_values`
basic_table() %>%
  count_values("Species", values = "setosa") %>%
  build_table(iris)

# `s_count_values.character`
s_count_values(x = c("a", "b", "a"), values = "a")
s_count_values(x = c("a", "b", "a", NA, NA), values = "b", na.rm = FALSE)

# `s_count_values.factor`
s_count_values(x = factor(c("a", "b", "a")), values = "a")

# `s_count_values.logical`
s_count_values(x = c(TRUE, FALSE, TRUE))

# `a_count_values`
a_count_values(x = factor(c("a", "b", "a")), values = "a", .N_col = 10, .N_row = 10)
```

cox_regression

Cox proportional hazards regression

Description

[Stable]

Fits a Cox regression model and estimates hazard ratio to describe the effect size in a survival analysis.
Usage

summarize_coxreg(
    lyt,
    variables,
    control = control_coxreg(),
    at = list(),
    multivar = FALSE,
    common_var = "STUDYID",
    .stats = c("n", "hr", "ci", "pval", "pval_inter"),
    .formats = c(n = "xx", hr = "xx.xx", ci = "(xx.xx, xx.xx)", pval = "x.xxxx | (<0.0001)", pval_inter = "x.xxxx | (<0.0001)"),
    varlabels = NULL,
    .indent_mods = NULL,
    na_str = "",
    .section_div = NA_character_
)

s_coxreg(model_df, .stats, .which_vars = "all", .var_nms = NULL)

a_coxreg(
    df,
    labelstr,
    eff = FALSE,
    var_main = FALSE,
    multivar = FALSE,
    variables,
    at = list(),
    control = control_coxreg(),
    .spl_context,
    .stats,
    .formats,
    .indent_mods = NULL,
    na_str = "",
    cache_env = NULL
)

Arguments

lyt                (PreDataTableLayouts)
layout that analyses will be added to.

variables          (named list of string)
list of additional analysis variables.

control            (list)
a list of parameters as returned by the helper function control_coxreg().

at                  (list of numeric)
when the candidate covariate is a numeric, use at to specify the value of the
covariate at which the effect should be estimated.
multivar (flag) whether multivariate Cox regression should run (defaults to FALSE), otherwise univariate Cox regression will run.

common_var (string) the name of a factor variable in the dataset which takes the same value for all rows. This should be created during pre-processing if no such variable currently exists.

.stats (character) the names of statistics to be reported among:
  • n: number of observations (univariate only)
  • hr: hazard ratio
  • ci: confidence interval
  • pval: p-value of the treatment effect
  • pval_inter: p-value of the interaction effect between the treatment and the covariate (univariate only)

.formats (named character or list) formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

varlabels (list) a named list corresponds to the names of variables found in data, passed as a named list and corresponding to time, event, arm, strata, and covariates terms. If arm is missing from variables, then only Cox model(s) including the covariates will be fitted and the corresponding effect estimates will be tabulated later.

.indent_mods (named integer) indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

na_str (string) custom string to replace all NA values with. Defaults to "".

.section_div (string or NA) string which should be repeated as a section divider between sections. Defaults to NA for no section divider. If a vector of two strings are given, the first will be used between treatment and covariate sections and the second between different covariates.

model_df (data.frame) contains the resulting model fit from a fit_coxreg function with tidying applied via broom::tidy().

.which_vars (character) which rows should statistics be returned for from the given model. Defaults to "all". Other options include "var_main" for main effects, "inter" for interaction effects, and "multi_lvl" for multivariate model covariate level rows. When .which_vars is "all", specific variables can be selected by specifying .var_nms.

.var_nms (character) the term value of rows in df for which .stats should be returned. Typically this is the name of a variable. If using variable labels, var should be a vector.
of both the desired variable name and the variable label in that order to see all stats related to that variable. When .which_vars is "var_main", .var_nms should be only the variable name.

df (data.frame)
data set containing all analysis variables.

labelstr (string)
label of the level of the parent split currently being summarized (must be present as second argument in Content Row Functions). See rtables::summarize_row_groups() for more information.

eff (flag)
whether treatment effect should be calculated. Defaults to FALSE.

var_main (flag)
whether main effects should be calculated. Defaults to FALSE.

.spl_context (data.frame)
gives information about ancestor split states that is passed by rtables.

cache_env (environment)
an environment object used to cache the regression model in order to avoid repeatedly fitting the same model for every row in the table. Defaults to NULL (no caching).

Details

Cox models are the most commonly used methods to estimate the magnitude of the effect in survival analysis. It assumes proportional hazards: the ratio of the hazards between groups (e.g., two arms) is constant over time. This ratio is referred to as the "hazard ratio" (HR) and is one of the most commonly reported metrics to describe the effect size in survival analysis (NEST Team, 2020).

Value

- summarize_coxreg() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add a Cox regression table containing the chosen statistics to the table layout.

- s_coxreg() returns the selected statistic for from the Cox regression model for the selected variable(s).

- a_coxreg() returns formatted rtables::CellValue().

Functions

- summarize_coxreg(): Layout-creating function which creates a Cox regression summary table layout. This function is a wrapper for several rtables layouting functions. This function is a wrapper for rtables::analyze_colvars() and rtables::summarize_row_groups().

- s_coxreg(): Statistics function that transforms results tabulated from fit_coxreg_univar() or fit_coxreg_multivar() into a list.

- a_coxreg(): Analysis function which is used as afun in rtables::analyze() and cfun in rtables::summarize_row_groups() within summarize_coxreg().
See Also

`fit_coxreg` for relevant fitting functions, `h_cox_regression` for relevant helper functions, and `tidy_coxreg` for custom tidy methods.

`fit_coxreg_univar()` and `fit_coxreg_multivar()` which also take the variables, data, at (univariate only), and control arguments but return unformatted univariate and multivariate Cox regression models, respectively.

Examples

```r
library(survival)

# Testing dataset [survival::bladder].
set.seed(1, kind = "Mersenne-Twister")
dta_bladder <- with(
  data = bladder[bladder$enum < 5, ],
  tibble::tibble(
    TIME = stop,
    STATUS = event,
    ARM = as.factor(rx),
    COVAR1 = as.factor(enum) %>% formatters::with_label("A Covariate Label"),
    COVAR2 = factor(
      sample(as.factor(enum)),
      levels = 1:4, labels = c("F", "F", "M", "M")
    ) %>% formatters::with_label("Sex (F/M)")
  )
)
dta_bladder$AGE <- sample(20:60, size = nrow(dta_bladder), replace = TRUE)
dta_bladder$STUDYID <- factor("X")

u1_variables <- list(
  time = "TIME", event = "STATUS", arm = "ARM", covariates = c("COVAR1", "COVAR2")
)

u2_variables <- list(time = "TIME", event = "STATUS", covariates = c("COVAR1", "COVAR2"))

m1_variables <- list(time = "TIME", event = "STATUS", arm = "ARM", covariates = c("COVAR1", "COVAR2")
)

m2_variables <- list(time = "TIME", event = "STATUS", covariates = c("COVAR1", "COVAR2"))

# summarize_coxreg
result_univar <- basic_table() %>%
  summarize_coxreg(variables = u1_variables) %>%
  build_table(dta_bladder)
result_univar

result_univar_covs <- basic_table() %>%
  summarize_coxreg(
    variables = u2_variables,
  )
```

```r

# Cox Regression

# Univariate
univar_model <- fit_coxreg_univar(variables = u1_variables, data = dta_bladder)
df1 <- broom::tidy(univar_model)
s_coxreg(model_df = df1, .stats = "hr")

# Univariate with interactions
univar_model_inter <- fit_coxreg_univar(
  variables = u1_variables, control = control_coxreg(interaction = TRUE), data = dta_bladder
)
df1_inter <- broom::tidy(univar_model_inter)
s_coxreg(model_df = df1_inter, .stats = "hr", .which_vars = "inter", .var_nms = "COVAR1")

# Univariate without treatment arm - only "COVAR2" covariate effects
univar_covs_model <- fit_coxreg_univar(variables = u2_variables, data = dta_bladder)
df1_covs <- broom::tidy(univar_covs_model)
s_coxreg(model_df = df1_covs, .stats = "hr", .var_nms = c("COVAR2", "Sex (F/M")")

# Multivariate.
multivar_model <- fit_coxreg_multivar(variables = m1_variables, data = dta_bladder)
df2 <- broom::tidy(multivar_model)
s_coxreg(model_df = df2, .stats = "pval", .which_vars = "var_main", .var_nms = "COVAR1")
s_coxreg(model_df = df2, .stats = "pval", .which_vars = "multi_lvl",
  .var_nms = c("COVAR1", "A Covariate Label")
)
# Multivariate without treatment arm - only "COVAR1" main effect
multivar_covs_model <- fit_coxreg_multivar(variables = m2_variables, data = dta_bladder)
df2_covs <- broom::tidy(multivar_covs_model)

s_coxreg(model_df = df2_covs, .stats = "hr")

a_coxreg(
  df = dta_bladder,
  labelstr = "Label 1",
  variables = u1_variables,
  .spl_context = list(value = "COVAR1"),
  .stats = "n",
  .formats = "xx"
)

a_coxreg(
  df = dta_bladder,
  labelstr = "",
  variables = u1_variables,
  .spl_context = list(value = "COVAR2"),
  .stats = "pval",
  .formats = "xx.xxxx"
)

---

**cox_regression_inter**

*Cox regression helper function for interactions*

**Description**

[Stable]

Test and estimate the effect of a treatment in interaction with a covariate. The effect is estimated as the HR of the tested treatment for a given level of the covariate, in comparison to the treatment control.

**Usage**

h_coxreg_inter_effect(x, effect, covar, mod, label, control, ...)

## S3 method for class 'numeric'
h_coxreg_inter_effect(x, effect, covar, mod, label, control, at, ...)

## S3 method for class 'factor'
h_coxreg_inter_effect(x, effect, covar, mod, label, control, data, ...)

## S3 method for class 'character'
h_coxreg_inter_effect(x, effect, covar, mod, label, control, data, ...)

h_coxreg_extract_interaction(effect, covar, mod, data, at, control)
h_coxreg_inter_estimations(
  variable,
  given,
  lvl_var,
  lvl_given,
  mod,
  conf_level = 0.95
)

Arguments

- **x** *(numeric or factor)*
  the values of the covariate to be tested.

- **effect** *(string)*
  the name of the effect to be tested and estimated.

- **covar** *(string)*
  the name of the covariate in the model.

- **mod** *(coxph)*
  a fitted Cox regression model (see `survival::coxph()`).

- **label** *(string)*
  the label to be returned as `term_label`.

- **control** *(list)*
  a list of controls as returned by `control_coxreg()`.

- **at** *(list)*
  a list with items named after the covariate, every item is a vector of levels at which the interaction should be estimated.

- **data** *(data.frame)*
  the data frame on which the model was fit.

- **variable, given** *(string)*
  the name of variables in interaction. We seek the estimation of the levels of `variable` given the levels of `given`.

- **lvl_var, lvl_given** *(character)*
  corresponding levels as given by `levels()`.

- **conf_level** *(proportion)*
  confidence level of the interval.

Details

Given the cox regression investigating the effect of Arm (A, B, C; reference A) and Sex (F, M; reference Female) and the model being abbreviated: `y ~ Arm + Sex + Arm:Sex`. The cox regression estimates the coefficients along with a variance-covariance matrix for:

- `b1` (arm b), `b2` (arm c)
The estimation of the Hazard Ratio for arm C/sex M is given in reference to arm A/sex M by \( \exp(b2 + b3 + b5)/ \exp(b3) = \exp(b2 + b5) \). The interaction coefficient is deduced by \( b2 + b5 \) while the standard error is obtained as \$\sqrt{\text{Var } b2 + \text{Var } b5 + 2 \times \text{covariance } (b2,b5)}\$.

**Value**

- \( \text{h_coxreg_inter_effect()} \) returns a data.frame of covariate interaction effects consisting of the following variables: effect, term, term_label, level, n, hr, lcl, ucl, pval, and pval_inter.

- \( \text{h_coxreg_extract_interaction()} \) returns the result of an interaction test and the estimated values. If no interaction, \( \text{h_coxreg_univar_extract()} \) is applied instead.

- \( \text{h_coxreg_inter_estimations()} \) returns a list of matrices (one per level of variable) with rows corresponding to the combinations of variable and given, with columns:
  - coef_hat: Estimation of the coefficient.
  - coef_se: Standard error of the estimation.
  - hr: Hazard ratio.
  - lcl, ucl: Lower/upper confidence limit of the hazard ratio.

**Functions**

- \( \text{h_coxreg_inter_effect()} \): S3 generic helper function to determine interaction effect.

- \( \text{h_coxreg_inter_effect(numeric)} \): Method for numeric class. Estimates the interaction with a numeric covariate.

- \( \text{h_coxreg_inter_effect(factor)} \): Method for factor class. Estimate the interaction with a factor covariate.

- \( \text{h_coxreg_inter_effect(character)} \): Method for character class. Estimate the interaction with a character covariate. This makes an automatic conversion to factor and then forwards to the method for factors.

- \( \text{h_coxreg_extract_interaction()} \): A higher level function to get the results of the interaction test and the estimated values.

- \( \text{h_coxreg_inter_estimations()} \): Hazard ratio estimation in interactions.

**Note**

- Automatic conversion of character to factor does not guarantee results can be generated correctly. It is therefore better to always pre-process the dataset such that factors are manually created from character variables before passing the dataset to \( \text{rtables::build_table()} \).
Examples

library(survival)

set.seed(1, kind = "Mersenne-Twister")

# Testing dataset [survival::bladder].
dta_bladder <- with(
data = bladder[bladder$enum < 5, ],
data.frame(
time = stop,
status = event,
armcd = as.factor(rx),
covar1 = as.factor(enum),
covar2 = factor(
sample(as.factor(enum)),
levels = 1:4,
labels = c("F", "F", "M", "M")
))
)
labels <- c("armcd" = "ARM", "covar1" = "A Covariate Label", "covar2" = "Sex (F/M)"

formatters::var_labels(dta_bladder)[names(labels)] <- labels
dta_bladder$age <- sample(20:60, size = nrow(dta_bladder), replace = TRUE)

plot(
  survfit(Surv(time, status) ~ armcd + covar1, data = dta_bladder),
  lty = 2:4,
xlab = "Months",
col = c("blue1", "blue2", "blue3", "blue4", "red1", "red2", "red3", "red4")
)

mod <- coxph(Surv(time, status) ~ armcd * covar1, data = dta_bladder)

h_coxreg_extract_interaction(
  mod = mod, effect = "armcd", covar = "covar1", data = dta_bladder,
  control = control_coxreg()
)

mod <- coxph(Surv(time, status) ~ armcd * covar1, data = dta_bladder)
result <- h_coxreg_inter_estimations(
  variable = "armcd", given = "covar1",
  lvl_var = levels(dta_bladder$armcd),
  lvl_given = levels(dta_bladder$covar1),
  mod = mod, conf_level = .95
)
result

cut_quantile_bins

Cut numeric vector into empirical quantile bins
cut_quantile_bins

Description

[Stable]
This cuts a numeric vector into sample quantile bins.

Usage

```r
cut_quantile_bins(
  x,
  probs = c(0.25, 0.5, 0.75),
  labels = NULL,
  type = 7,
  ordered = TRUE
)
```

Arguments

- **x**  
  (numeric)
  the continuous variable values which should be cut into quantile bins. This may contain NA values, which are then not used for the quantile calculations, but included in the return vector.

- **probs**  
  (numeric)
  the probabilities identifying the quantiles. This is a sorted vector of unique proportion values, i.e. between 0 and 1, where the boundaries 0 and 1 must not be included.

- **labels**  
  (character)
  the unique labels for the quantile bins. When there are \( n \) probabilities in `probs`, then this must be \( n + 1 \) long.

- **type**  
  (integer(1))
  type of quantiles to use, see `stats::quantile()` for details.

- **ordered**  
  (flag)
  should the result be an ordered factor.

Value

A factor variable with appropriately-labeled bins as levels.

Note

Intervals are closed on the right side. That is, the first bin is the interval \([-\infty, q_1]\) where \(q_1\) is the first quantile, the second bin is then \((q_1, q_2]\), etc., and the last bin is \((q_n, +\infty]\) where \(q_n\) is the last quantile.

Examples

```r
# Default is to cut into quartile bins.
cut_quantile_bins(cars$speed)

# Use custom quantiles.
```
cut_quantile_bins(cars$speed, probs = c(0.1, 0.2, 0.6, 0.88))

# Use custom labels.
cut_quantile_bins(cars$speed, labels = paste0("Q", 1:4))

# NAs are preserved in result factor.
ozone_binned <- cut_quantile_bins(airquality$Ozone)
which(is.na(ozone_binned))
# So you might want to make these explicit.
explicit_na(ozone_binned)

---

## day2month

**Conversion of days to months**

### Description

Conversion of days to months

### Usage

```r
day2month(x)
```

### Arguments

- `x` (numeric(1))
  - time in days.

### Value

A numeric vector with the time in months.

### Examples

```r
x <- c(403, 248, 30, 86)
day2month(x)
```
Add titles, footnotes, page Number, and a bounding box to a grid grob

Description

[Stable]
This function is useful to label grid grobs (also ggplot2, and lattice plots) with title, footnote, and page numbers.

Usage

```r
decorate_grob(
  grob,
  titles,
  footnotes,
  page = "",
  width_titles = grid::unit(1, "npc"),
  width_footnotes = grid::unit(1, "npc"),
  border = TRUE,
  padding = grid::unit(rep(1, 4), "lines"),
  margins = grid::unit(c(1, 0, 1, 0), "lines"),
  outer_margins = grid::unit(c(2, 1.5, 3, 1.5), "cm"),
  gp_titles = grid::gpar(),
  gp_footnotes = grid::gpar(fontsize = 8),
  name = NULL,
  gp = grid::gpar(),
  vp = NULL
)
```

Arguments

grob (grob)
a grid grob object, optionally NULL if only a grob with the decoration should be shown.

titles (character)
titles given as a vector of strings that are each separated by a newline and wrapped according to the page width.

footnotes (character)
footnotes. Uses the same formatting rules as titles.

page (string or NULL)
page numeration. If NULL then no page number is displayed.

width_titles (grid::unit)
width of titles. Usually defined as all the available space grid::unit(1, "npc"), it is affected by the parameter outer_margins. Right margins (outer_margins[4]) need to be subtracted to the allowed width.
width_footnotes
  (grid::unit)
  width of footnotes. Same default and margin correction as width_titles.

border
  (flag)
  whether a border should be drawn around the plot or not.

padding
  (grid::unit)
  padding. A unit object of length 4. Innermost margin between the plot (grob) and, possibly, the border of the plot. Usually expressed in 4 identical values (usually "lines"). It defaults to grid::unit(rep(1, 4), "lines").

margins
  (grid::unit)
  margins. A unit object of length 4. Margins between the plot and the other elements in the list (e.g. titles, plot, and footers). This is usually expressed in 4 "lines", where the lateral ones are 0s, while top and bottom are 1s. It defaults to grid::unit(c(1, 0, 1, 0), "lines").

outer_margins
  (grid::unit)
  outer margins. A unit object of length 4. It defines the general margin of the plot, considering also decorations like titles, footnotes, and page numbers. It defaults to grid::unit(c(2, 1.5, 3, 1.5), "cm").

gp_titles
  (gpar)
  a gpar object. Mainly used to set different "fontsize".

gp_footnotes
  (gpar)
  a gpar object. Mainly used to set different "fontsize".

name
  a character identifier for the grob. Used to find the grob on the display list and/or as a child of another grob.

gp
  A "gpar" object, typically the output from a call to the function gpar. This is basically a list of graphical parameter settings.

vp
  a viewport object (or NULL).

Details
  The titles and footnotes will be ragged, i.e. each title will be wrapped individually.

Value
  A grid grob (gTree).

Examples
  library(grid)

  titles <- c(
    "Edgar Anderson's Iris Data",
    paste(
      "This famous (Fisher's or Anderson's) iris data set gives the measurements",
      "in centimeters of the variables sepal length and width and petal length",
      "and width, respectively, for 50 flowers from each of 3 species of iris."
    )
  )
decorate_grob

footnotes <- c(
  "The species are Iris setosa, versicolor, and virginica."
  paste(
    "iris is a data frame with 150 cases (rows) and 5 variables (columns) named",
  ),
)

## empty plot
grid.newpage()

grid.draw(
  decorate_grob(
    NULL,
    titles = titles,
    footnotes = footnotes,
    page = "Page 4 of 10"
  )
)

# grid
p <- gTree(
  children = gList(
    rectGrob(),
    xaxisGrob(),
    yaxisGrob(),
    textGrob("Sepal.Length", y = unit(-4, "lines")),
    textGrob("Petal.Length", x = unit(-3.5, "lines"), rot = 90),
    pointsGrob(iris$Sepal.Length, iris$Petal.Length, gp = gpar(col = iris$Species), pch = 16)
  ),
  vp = vpStack(plotViewport(), dataViewport(xData = iris$Sepal.Length, yData = iris$Petal.Length))
)
grid.newpage()
grid.draw(p)

grid.newpage()
grid.draw(
  decorate_grob(
    grob = p,
    titles = titles,
    footnotes = footnotes,
    page = "Page 6 of 129"
  )
)

## with ggplot2
library(ggplot2)
p_gg <- ggplot2::ggplot(iris, aes(Sepal.Length, Sepal.Width, col = Species)) +
ggplot2::geom_point()
p_gg
decorate_grob_set

Decorate set of grobs and add page numbering

Description

[Stable]

Note that this uses the decorate_grob_factory() function.
Usage

decorate_grob_set(grobs, ...)

Arguments

grobs (list of grob)  
a list of grid grobs.
...
arguments passed on to decorate_grob().

Value

A decorated grob.

Examples

library(ggplot2)
library(grid)
g <- with(data = iris, {
  list(
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Sepal.Length, Sepal.Width, col = Species)) +
      ggplot2::geom_point()
    ),
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Sepal.Length, Petal.Length, col = Species)) +
      ggplot2::geom_point()
    ),
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Sepal.Length, Petal.Width, col = Species)) +
      ggplot2::geom_point()
    ),
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Sepal.Width, Petal.Length, col = Species)) +
      ggplot2::geom_point()
    ),
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Sepal.Width, Petal.Width, col = Species)) +
      ggplot2::geom_point()
    ),
    ggplot2::ggplotGrob(
      ggplot2::ggplot(mapping = aes(Petal.Length, Petal.Width, col = Species)) +
      ggplot2::geom_point()
    )
  )
})
lg <- decorate_grob_set(grobs = g, titles = "Hello\nOne\nTwo\nThree", footnotes = "")
draw_grob(lg[[1]])
draw_grob(lg[[2]])
draw_grob(lg[[6]])
default_na_str Default string replacement for NA values

Description

[Stable]
The default string used to represent NA values. This value is used as the default value for the na_str argument throughout the tern package, and printed in place of NA values in output tables. If not specified for each tern function by the user via the na_str argument, or in the R environment options via set_default_na_str(), then NA is used.

Usage

default_na_str()

set_default_na_str(na_str)

Arguments

na_str (string)
   single string value to set in the R environment options as the default value to replace NAs. Use getOption("tern_default_na_str") to check the current value set in the R environment (defaults to NULL if not set).

Value

  - default_na_str returns the current value if an R environment option has been set for "tern_default_na_str", or NA_character_ otherwise.
  - set_default_na_str has no return value.

Functions

  - default_na_str(): Accessor for default NA value replacement string.
  - set_default_na_str(): Setter for default NA value replacement string. Sets the option "tern_default_na_str" within the R environment.

Examples

# Default settings
default_na_str()
getOption("tern_default_na_str")

# Set custom value
set_default_na_str("<Missing>")

# Settings after value has been set
default_na_str()
default_stats_formats_labels

Get default statistical methods and their associated formats, labels, and indent modifiers

Description

[Experimental]
Utility functions to get valid statistic methods for different method groups (.stats) and their associated formats (.formats), labels (.labels), and indent modifiers (.indent_mods). This utility is used across tern, but some of its working principles can be seen in analyze_vars(). See notes to understand why this is experimental.

Usage

get_stats(
  method_groups = "analyze_vars_numeric",
  stats_in = NULL,
  add_pval = FALSE
)

get_formats_from_stats(stats, formats_in = NULL)

get_labels_from_stats(stats, labels_in = NULL, row_nms = NULL)

get_indents_from_stats(stats, indents_in = NULL, row_nms = NULL)

tern_default_stats

tern_default_formats

tern_default_labels

summary_formats(type = "numeric", include_pval = FALSE)

summary_labels(type = "numeric", include_pval = FALSE)

Arguments

method_groups (character)
indicates the statistical method group (tern_analyze function) to retrieve default statistics for. A character vector can be used to specify more than one statistical method group.
default_stats_formats_labels

stats_in (character)
statistics to retrieve for the selected method group.

add_pval (flag)
should "pval" (or "pval_counts" if method_groups contains "analyze-vars_counts")
be added to the statistical methods?

stats (character)
statistical methods to get defaults for.

formats_in (named vector)
inserted formats to replace defaults. It can be a character vector from formatters::list_valid_format_labels()
or a custom format function.

labels_in (named character)
inserted labels to replace defaults.

row_nms (character)
row names. Levels of a factor or character variable, each of which the statistics in .stats will be calculated for. If this parameter is set, these variable levels will be used as the defaults, and the names of the given custom values should correspond to levels (or have format statistic.level) instead of statistics. Can also be variable names if rows correspond to different variables instead of levels. Defaults to NULL.

indents_in (named vector)
inserted indent modifiers to replace defaults (default is 0L).

type (string)
"numeric" or "counts".

include_pval (flag)
same as the add_pval argument in get_stats().

Format

- tern_default_stats is a named list of available statistics, with each element named for their corresponding statistical method group.

- tern_default_formats is a named vector of available default formats, with each element named for their corresponding statistic.

- tern_default_labels is a named character vector of available default labels, with each element named for their corresponding statistic.

Details

Current choices for type are counts and numeric for analyze-vars() and affect get_stats().

Value

- get_stats() returns a character vector of statistical methods.

- get_formats_from_stats() returns a named vector of formats (if present in either tern_default_formats or formats_in, otherwise NULL). Values can be taken from formatters::list_valid_format_labels() or a custom function (e.g. formatting_functions).
• `get_labels_from_stats()` returns a named character vector of labels (if present in either `tern_default_labels` or `labels_in`, otherwise NULL).

• `get_indents_from_stats()` returns a single indent modifier value to apply to all rows or a named numeric vector of indent modifiers (if present, otherwise NULL).

• `summary_formats()` returns a named vector of default statistic formats for the given data type.

• `summary_labels` returns a named vector of default statistic labels for the given data type.

Functions
• `get_stats()`: Get statistics available for a given method group (analyze function).
• `get_formats_from_stats()`: Get formats corresponding to a list of statistics.
• `get_labels_from_stats()`: Get labels corresponding to a list of statistics.
• `get_indents_from_stats()`: Format indent modifiers for a given vector/list of statistics.
• `tern_default_stats`: Named list of available statistics by method group for tern.
• `tern_default_formats`: Named vector of default formats for tern.
• `tern_default_labels`: Named character vector of default labels for tern.
• `summary_formats()`: Quick function to retrieve default formats for summary statistics: `analyze_vars()` and `analyze_vars_in_cols()` principally.
• `summary_labels()`: Quick function to retrieve default labels for summary statistics. Returns labels of descriptive statistics which are understood by rtables. Similar to `summary_formats`.

Note
These defaults are experimental because we use the names of functions to retrieve the default statistics. This should be generalized in groups of methods according to more reasonable groupings.
Formats in `tern` and `rtables` can be functions that take in the table cell value and return a string. This is well documented in vignette("custom_appearance", package = "rtables").

See Also
`formatting_functions`

Examples

```r
# analyze_vars is numeric
num_stats <- get_stats("analyze_vars_numeric") # also the default

# Other type
cnt_stats <- get_stats("analyze_vars_counts")

# Weirdly taking the pval from count_occurrences
only_pval <- get_stats("count_occurrences", add_pval = TRUE, stats_in = "pval")

# All count_occurrences
```
all_cnt_occ <- get_stats("count_occurrences")

# Multiple
get_stats(c("count_occurrences", "analyze_vars_counts"))

# Defaults formats
get_formats_from_stats(num_stats)
get_formats_from_stats(cnt_stats)
get_formats_from_stats(only_pval)
get_formats_from_stats(all_cnt_occ)

# Addition of customs
get_formats_from_stats(all_cnt_occ, formats_in = c("fraction" = c("xx")))
get_formats_from_stats(all_cnt_occ, formats_in = list("fraction" = c("xx.xx", "xx")))

# Defaults labels
get_labels_from_stats(num_stats)
get_labels_from_stats(cnt_stats)
get_labels_from_stats(only_pval)
get_labels_from_stats(all_cnt_occ)

# Addition of customs
get_labels_from_stats(all_cnt_occ, labels_in = c("fraction" = "Fraction"))
get_labels_from_stats(all_cnt_occ, labels_in = list("fraction" = c("Some more fractions")))

get_indents_from_stats(all_cnt_occ, indents_in = 3L)
get_indents_from_stats(all_cnt_occ, indents_in = list(count = 2L, count_fraction = 5L))
get_indents_from_stats(all_cnt_occ,
  indents_in = list(a = 2L, count.a = 1L, count.b = 5L), row_nms = c("a", "b")
)

summary_formats()
summary_formats(type = "counts", include_pval = TRUE)

summary_labels()
summary_labels(type = "counts", include_pval = TRUE)

---

**df_explicit_na**

Encode categorical missing values in a data frame

**Description**

[Stable]

This is a helper function to encode missing entries across groups of categorical variables in a data frame.
Usage

df_explicit_na(
  data,
  omit_columns = NULL,
  char_as_factor = TRUE,
  logical_as_factor = FALSE,
  na_level = "<Missing>"
)

Arguments

data (data.frame)
data set.

omit_columns (character)
names of variables from data that should not be modified by this function.

char_as_factor (flag)
whether to convert character variables in data to factors.

logical_as_factor (flag)
whether to convert logical variables in data to factors.

na_level (string)
string used to replace all NA or empty values inside non-omit_columns columns.

Details

Missing entries are those with NA or empty strings and will be replaced with a specified value. If factor variables include missing values, the missing value will be inserted as the last level. Similarly, in case character or logical variables should be converted to factors with the char_as_factor or logical_as_factor options, the missing values will be set as the last level.

Value

A data.frame with the chosen modifications applied.

See Also

sas_na() and explicit_na() for other missing data helper functions.

Examples

my_data <- data.frame(
  u = c(TRUE, FALSE, NA, TRUE),
  v = factor(c("A", NA, NA, NA), levels = c("Z", "A")),
  w = c("A", "B", NA, "C"),
  x = c("D", "E", "F", NA),
  y = c("G", "H", "I", ""),
  z = c(1, 2, 3, 4),
  stringsAsFactors = FALSE
)
# Example 1
# Encode missing values in all character or factor columns.
df_explicit_na(my_data)
# Also convert logical columns to factor columns.
df_explicit_na(my_data, logical_as_factor = TRUE)
# Encode missing values in a subset of columns.
df_explicit_na(my_data, omit_columns = c("x", "y"))

# Example 2
# Here we purposefully convert all 'M' values to 'NA' in the 'SEX' variable.
# After running `df_explicit_na` the 'NA' values are encoded as '<Missing>' but they are not
# included when generating 'rtables'.
adsl <- tern_ex_adsl
adsl$SEX[adsl$SEX == "M"] <- NA
adsl <- df_explicit_na(adsl)

# If you want the 'Na' values to be displayed in the table use the 'na_level' argument.
adsl <- tern_ex_adsl
adsl$SEX[adsl$SEX == "M"] <- NA
adsl <- df_explicit_na(adsl, na_level = "Missing Values")

# Example 3
# Numeric variables that have missing values are not altered. This means that any 'NA' value in
# a numeric variable will not be included in the summary statistics, nor will they be included
# in the denominator value for calculating the percent values.
adsl <- tern_ex_adsl
adsl$AGE[adsl$AGE < 30] <- NA
adsl <- df_explicit_na(adsl)

---

draw_grob Draw grob

Description

[Deprecated]
Draw grob on device page.

Usage

draw_grob(grob, newpage = TRUE, vp = NULL)

Arguments

grob (grob) grid object.
newpage (flag) draw on a new page.
value

A grob.

Examples

library(dplyr)
library(grid)

rect <- rectGrob(width = grid::unit(0.5, "npc"), height = grid::unit(0.5, "npc"))
rect %>% draw_grob(vp = grid::viewport(angle = 45))

num <- lapply(1:10, textGrob)
num %>%
  arrange_grobs(grobs = .) %>%
  draw_grob()
showViewport()
**d_count_cumulative**

Description

[Stable]

This is a helper function that describes the analysis in `s_count_cumulative()`.

Usage

d_count_cumulative(threshold, lower_tail, include_eq)

Arguments

- **threshold** (numeric(1))
  
a cutoff value as threshold to count values of x.
- **lower_tail** (flag)
  
whether to count lower tail, default is TRUE.
- **include_eq** (flag)
  
whether to include value equal to the threshold in count, default is TRUE.

Value

Labels for `s_count_cumulative()`.

**d_count_missed_doses**

Description

[Stable]

Usage

d_count_missed_doses(thresholds)

Arguments

- **thresholds** (numeric)
  
minimum number of missed doses the patients had.

Value

`d_count_missed_doses()` returns a named character vector with the labels.
d_onco_rsp_label

Description

[Stable]
Describe the oncology response in a standard way.

Usage

d_onco_rsp_label(x)

Arguments

x (character)
the standard oncology codes to be described.

Value

Response labels.

See Also

estimate_multinomial_rsp()

Examples

d_onco_rsp_label(
)

# Adding some values not considered in d_onco_rsp_label

d_onco_rsp_label(
  c("CR", "PR", "hello", "hi")
)
**d_pkparam**  
*Generate PK reference dataset*

**Description**

[Stable]

**Usage**

```r
d_pkparam()
```

**Value**

A `data.frame` of PK parameters.

**Examples**

```r
pk_reference_dataset <- d_pkparam()
```

---

**d_proportion**  
*Description of the proportion summary*

**Description**

[Stable]

This is a helper function that describes the analysis in `s_proportion()`.

**Usage**

```r
d_proportion(conf_level, method, long = FALSE)
```

**Arguments**

- `conf_level` *(proportion)*  
  confidence level of the interval.

- `method` *(string)*  
  the method used to construct the confidence interval for proportion of successful outcomes; one of `waldcc`, `wald`, `clopper-pearson`, `wilson`, `wilsonc`, `strat_wilson`, `strat_wilsonc`, `agresti-coull`, or `jeffreys`.

- `long` *(flag)*  
  whether a long or a short (default) description is required.

**Value**

String describing the analysis.
**d_proportion_diff**  
*Description of method used for proportion comparison*

---

**Description**  

[Stable]

This is an auxiliary function that describes the analysis in `s_proportion_diff()`.

**Usage**

```r
d_proportion_diff(conf_level, method, long = FALSE)
```

**Arguments**

- `conf_level` *(proportion)*  
  confidence level of the interval.
- `method` *(string)*  
  the method used for the confidence interval estimation.
- `long` *(flag)*  
  whether a long (TRUE) or a short (FALSE, default) description is required.

**Value**

A string describing the analysis.

**See Also**

`prop_diff`

---

**d_rsp_subgroups_colvars**  
*Labels for column variables in binary response by subgroup table*

---

**Description**  

[Stable]

Internal function to check variables included in `tabulate_rsp_subgroups()` and create column labels.

**Usage**

```r
d_rsp_subgroups_colvars(vars, conf_level = NULL, method = NULL)
```
**Arguments**

- **vars** (character)
  variable names for the primary analysis variable to be iterated over.

- **conf_level** (proportion)
  confidence level of the interval.

- **method** (string or NULL)
  specifies the test used to calculate the p-value for the difference between two proportions. For options, see `s_test_proportion_diff()`. Default is NULL so no test is performed.

**Value**

A list of variables to tabulate and their labels.

---

**Description**

[Stable]

Internal function to check variables included in `tabulate_survival_subgroups()` and create column labels.

**Usage**

```r
d_survival_subgroups_colvars(vars, conf_level, method, time_unit = NULL)
```

**Arguments**

- **vars** (character)
  the names of statistics to be reported among:
  
  - n_tot_events: Total number of events per group.
  - n_events: Number of events per group.
  - n_tot: Total number of observations per group.
  - n: Number of observations per group.
  - median: Median survival time.
  - hr: Hazard ratio.
  - ci: Confidence interval of hazard ratio.
  - pval: p-value of the effect. Note, one of the statistics n_tot and n_tot_events, as well as both hr and ci are required.

- **conf_level** (proportion)
  confidence level of the interval.
method   (string)
p-value method for testing hazard ratio = 1.

time_unit (string)
label with unit of median survival time. Default NULL skips displaying unit.

Value
A list of variables and their labels to tabulate.

Note
At least one of n.tot and n.tot_events must be provided in vars.

---

d_test_proportion_diff

Description of the difference test between two proportions

Description

[Stable]
This is an auxiliary function that describes the analysis in s_test_proportion_diff.

Usage
d_test_proportion_diff(method)

Arguments

method   (string)
one of chisq, cmh, fisher, or schouten; specifies the test used to calculate the p-value.

Value
A string describing the test from which the p-value is derived.
**Description**

[Stable]

Estimate the proportion along with confidence interval of a proportion regarding the level of a factor.

**Usage**

```r
estimate_multinomial_response(
  lyt,
  var,
  na_str = default_na_str(),
  nested = TRUE,
  ...
)
```

```r
s_length_proportion(x, .N_col, ...)
```

```r
a_length_proportion(x, .N_col, ...)
```

**Arguments**

- **lyt** *(PreDataTableLayouts)*
  layout that analyses will be added to.

- **var** *(string)*
  single variable name that is passed by `rtables` when requested by a statistics function.

- **na_str** *(string)*
  string used to replace all NA or empty values in the output.

- **nested** *(flag)*
  whether this layout instruction should be applied within the existing layout structure _if possible_ (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split underneath analyses, which is not allowed.

- **...**
  additional arguments for the lower level functions.

- **show_labels** *(string)*
  label visibility: one of "default", "visible" and "hidden".
table_names (character)  
this can be customized in the case that the same vars are analyzed multiple 
times, to avoid warnings from rtables.

.stats (character)  
statistics to select for the table. Run get_stats("estimate_multinomial_response")
to see available statistics for this function.

.formats (named character or list)  
formats for the statistics. See Details in analyze_vars for more information on
the "auto" setting.

.labels (named character)  
labels for the statistics (without indent).

.indent_mods (named integer)  
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified
default behavior. Can be negative.

x (numeric)  
vector of numbers we want to analyze.

.N_col (integer(1))  
column-wise N (column count) for the full column being analyzed that is typi-
cally passed by rtables.

Value

- estimate_multinomial_response() returns a layout object suitable for passing to further
layouting functions, or to rtables::build_table(). Adding this function to an rtable
layout will add formatted rows containing the statistics from s_length_proportion() to the

table layout.

- s_length_proportion() returns statistics from s_proportion().

- a_length_proportion() returns the corresponding list with formatted rtables::CellValue().

Functions

- estimate_multinomial_response(): Layout-creating function which can take statistics func-
tion arguments and additional format arguments. This function is a wrapper for rtables::analyze()
and rtables::summarize_row_groups().

- s_length_proportion(): Statistics function which feeds the length of x as number of suc-
cesses, and .N_col as total number of successes and failures into s_proportion().

- a_length_proportion(): Formatted analysis function which is used as afun in estimate_multinomial_response().

See Also

Relevant description function d_onco_rsp_label().
**Examples**

```r
library(dplyr)

data(dta_test)
# Use of the layout creating function.
dta_test <- data.frame(
  USUBJID = paste0("S", 1:12),
  ARM = factor(rep(LETTERS[1:3], each = 4)),
  AVAL = c(A = c(1, 1, 1, 1), B = c(0, 0, 1, 1), C = c(0, 0, 0, 0))
) %>% mutate(
  AVALC = factor(AVAL,
    levels = c(0, 1),
    labels = c("Complete Response (CR)", "Partial Response (PR)")
)

lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  estimate_multinomial_response(var = "AVALC")

tbl <- build_table(lyt, dta_test)

tbl

s_length_proportion(rep("CR", 10), .N_col = 100)
s_length_proportion(factor(character(0)), .N_col = 100)

a_length_proportion(rep("CR", 10), .N_col = 100)
a_length_proportion(factor(character(0)), .N_col = 100)
```

---

**estimate_proportions** *Estimation of proportions*

**Description**

[Stable]

Estimate the proportion of responders within a studied population.

**Usage**

```r
estimate_proportion(
  lyt, 
  vars, 
  conf_level = 0.95, 
  method = c("waldcc", "wald", "clopper-pearson", "wilson", "wilsonc", "strat_wilson", 
    "strat_wilsonc", "agresti-coull", "jeffreys"), 
  weights = NULL, 
  max_iterations = 50, 
)```
variables = list(strata = NULL),
long = FALSE,
na_str = default_na_str(),
nested = TRUE,
...,  
show_labels = "hidden",
table_names = vars,
.stats = NULL,
.formats = NULL,
.labels = NULL,
.indent.mods = NULL
)

s_proportion(
  df,
  .var,
  conf_level = 0.95,
  method = c("waldcc", "wald", "clopper-pearson", "wilson", "wilsonc", "strat_wilson",
             "strat_wilsonc", "agresti-coull", "jeffreys"),
  weights = NULL,
  max_iterations = 50,
  variables = list(strata = NULL),
  long = FALSE
)

a_proportion(
  df,
  .var,
  conf_level = 0.95,
  method = c("waldcc", "wald", "clopper-pearson", "wilson", "wilsonc", "strat_wilson",
             "strat_wilsonc", "agresti-coull", "jeffreys"),
  weights = NULL,
  max_iterations = 50,
  variables = list(strata = NULL),
  long = FALSE
)

Arguments

lyt  (PreDataTableLayouts)
layout that analyses will be added to.

vars  (character)
variable names for the primary analysis variable to be iterated over.

conf_level  (proportion)
confidence level of the interval.

method  (string)
the method used to construct the confidence interval for proportion of successful outcomes; one of waldcc, wald, clopper-pearson, wilson, wilsonc,
strat_wilson, strat_wilsonc, agresti-coull or jeffreys.

weights (numeric or NULL)
weights for each level of the strata. If NULL, they are estimated using the iterative 
algorithm proposed in Yan and Su (2010) that minimizes the weighted squared 
length of the confidence interval.

max_iterations (count)
maximum number of iterations for the iterative procedure used to find estimates 
of optimal weights.

variables (named list of string)
list of additional analysis variables.

long (flag)
whether a long description is required.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout 
structure _if possible (TRUE, the default) or as a new top-level element (FALSE). 
Ignored if it would nest a split. underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

table_names (character)
this can be customized in the case that the same vars are analyzed multiple 
times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("estimate_proportion") to 
see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on 
the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodi-
fied default behavior. Can be negative.

df (logical or data.frame)
if only a logical vector is used, it indicates whether each subject is a responder or 
not. TRUE represents a successful outcome. If a data.frame is provided, also the 
strata variable names must be provided in variables as a list element with 
the strata strings. In the case of data.frame, the logical vector of responses 
must be indicated as a variable name in .var.

.var (string)
single variable name that is passed by rtables when requested by a statistics 
function.
Value

- `estimate_proportion()` returns a layout object suitable for passing to further layouting functions, or to `rtables::build_table()`. Adding this function to an rtable layout will add formatted rows containing the statistics from `s_proportion()` to the table layout.

- `s_proportion()` returns statistics `n_prop` (n and proportion) and `prop_ci` (proportion CI) for a given variable.

- `a_proportion()` returns the corresponding list with formatted `rtables::CellValue()`.

Functions

- `estimate_proportion()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.

- `s_proportion()`: Statistics function estimating a proportion along with its confidence interval.

- `a_proportion()`: Formatted analysis function which is used as `afun` in `estimate_proportion()`.

See Also

`h_proportions`

Examples

```r
# Data generation

dta_test <- data.frame(
  USUBJID = paste0("S", 1:12),
  ARM = rep(LETTERS[1:3], each = 4),
  AVAL = c(A = c(1, 1, 1, 1), B = c(0, 0, 1, 1), C = c(0, 0, 0, 0))
)

basic_table() %>%
  split_cols_by("ARM") %>%
  estimate_proportion(vars = "AVAL") %>%
  build_table(df = dta_test)

# Case with only logical vector.

rsp_v <- c(1, 0, 1, 0, 1, 1, 0, 0)
s_proportion(rsp_v)

# Example for Stratified Wilson CI

nex <- 100 # Number of example rows

dta <- data.frame(
  "rsp" = sample(c(TRUE, FALSE), nex, TRUE),
  "grp" = sample(c("A", "B"), nex, TRUE),
  "f1" = sample(c("a1", "a2"), nex, TRUE),
  "f2" = sample(c("x", "y", "z"), nex, TRUE),
  stringsAsFactors = TRUE
)

s_proportion(
  df = dta,
)
.var = "rsp",
variables = list(strata = c("f1", "f2"),
conf_level = 0.90,
method = "strat_wilson"
)

explicit_na

Missing data

Description

[Stable]
Substitute missing data with a string or factor level.

Usage

explicit_na(x, label = "<Missing>")

Arguments

x (factor or character)
values for which any missing values should be substituted.

label (string)
string that missing data should be replaced with.

Value

x with any NA values substituted by label.

Examples

explicit_na(c(NA, "a", "b"))
    is.na(explicit_na(c(NA, "a", "b")))

explicit_na(factor(c(NA, "a", "b")))
    is.na(explicit_na(factor(c(NA, "a", "b"))))

explicit_na(sas_na(c("a", "")))
extract_rsp_biomarkers

Prepare response data estimates for multiple biomarkers in a single data frame

Description

[Stable]
Prepares estimates for number of responses, patients and overall response rate, as well as odds ratio estimates, confidence intervals and p-values, for multiple biomarkers across population subgroups in a single data frame. variables corresponds to the names of variables found in data, passed as a named list and requires elements rsp and biomarkers (vector of continuous biomarker variables) and optionally covariates, subgroups and strata. groups_lists optionally specifies groupings for subgroups variables.

Usage

extract_rsp_biomarkers(
    variables,
    data,
    groups_lists = list(),
    control = control_logistic(),
    label_all = "All Patients"
)

Arguments

variables  (named list of string)
    list of additional analysis variables.
data       (data.frame)
    the dataset containing the variables to summarize.
groups_lists (named list of list)
    optionally contains for each subgroups variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.
control     (named list)
    controls for the response definition and the confidence level produced by control_logistic().
label_all   (string)
    label for the total population analysis.

Value

A data.frame with columns biomarker, biomarker_label, n_tot, n_rsp, prop, or, lcl, ucl, conf_level, pval, pval_label, subgroup, var, var_label, and row_type.
Note

You can also specify a continuous variable in rsp and then use the response_definition control to convert that internally to a logical variable reflecting binary response.

See Also

h_logistic_mult_cont_df() which is used internally.

Examples

library(dplyr)
library(forcats)

adrs <- tern_ex_adrs
adrs_labels <- formatters::var_labels(adrs)

adrs_f <- adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  mutate(rsp = AVALC == "CR")

# Typical analysis of two continuous biomarkers `BMRKR1` and `AGE`,
# in logistic regression models with one covariate `RACE`. The subgroups
# are defined by the levels of `BMRKR2`.
df <- extract_rsp_biomarkers(
  variables = list(
    rsp = "rsp",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
    subgroups = "BMRKR2"
  ),
  data = adrs_f
)

df

# Here we group the levels of `BMRKR2` manually, and we add a stratification
# variable `STRATA1`. We also here use a continuous variable `EOSDY`
# which is then binarized internally (response is defined as this variable
# being larger than 750).
df_grouped <- extract_rsp_biomarkers(
  variables = list(
    rsp = "EOSDY",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
    subgroups = "BMRKR2",
    strata = "STRATA1"
  ),
  data = adrs_f,
  groups_lists = list(
    BMRKR2 = list(
      "low" = "LOW",
      "low/medium" = c("LOW", "MEDIUM"),
      "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
    )
  )
)
extract_rsp_subgroups

Prepare response data for population subgroups in data frames

Description

[Stable]
Prepares response rates and odds ratios for population subgroups in data frames. Simple wrapper for `h_odds_ratio_subgroups_df()` and `h_proportion_subgroups_df()`. Result is a list of two data.frames: prop and or. variables corresponds to the names of variables found in data, passed as a named list and requires elements rsp, arm and optionally subgroups and strata. groups_lists optionally specifies groupings for subgroups variables.

Usage

```r
extract_rsp_subgroups(
  variables,  # (named list of string)
  data,  # (data.frame)
  groups_lists = list(),  # list of additional analysis variables.
  data,  # the dataset containing the variables to summarize.
  conf_level = 0.95,  # optionally contains for each subgroups variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.
  method = NULL,  # confidence level of the interval.
  label_all = "All Patients"
)
```

Arguments

- `variables` (named list of string)
- `data` (data.frame)
- `groups_lists` (named list of list)
- `conf_level` (proportion)
extract_survival_biomarkers

Prepare survival data estimates for multiple biomarkers in a single data frame

Description

[Stable]

Prepares estimates for number of events, patients and median survival times, as well as hazard ratio estimates, confidence intervals and p-values, for multiple biomarkers across population subgroups in a single data frame. variables corresponds to the names of variables found in data, passed as a named list and requires elements tte, is_event, biomarkers (vector of continuous biomarker variables), and optionally subgroups and strata. groups_lists optionally specifies groupings for subgroups variables.

Usage

extract_survival_biomarkers(
  variables,
  data,
  groups_lists = list(),
  control = control_coxreg(),
  label_all = "All Patients"
)
extract_survival_subgroups

**Arguments**

- **variables** (named list of string): list of additional analysis variables.
- **data** (data.frame): the dataset containing the variables to summarize.
- **groups_lists** (named list of list): optionally contains for each subgroups variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.
- **control** (list): a list of parameters as returned by the helper function `control_coxreg()`.
- **label_all** (string): label for the total population analysis.

**Value**

A data.frame with columns biomarker, biomarker_label, n_tot, n_tot_events, median, hr, lcl, ucl, conf_level, pval, pval_label, subgroup, var, var_label, and row_type.

**See Also**

- `h_coxreg_mult_cont_df()` which is used internally, `tabulate_survival_biomarkers()`.

---

**extract_survival_subgroups**

*Prepare survival data for population subgroups in data frames*

**Description**

[Stable]

Prepares estimates of median survival times and treatment hazard ratios for population subgroups in data frames. Simple wrapper for `h_survtime_subgroups_df()` and `h_coxph_subgroups_df()`. Result is a list of two data.frames: survtime and hr. variables corresponds to the names of variables found in data, passed as a named list and requires elements tte, is_event, arm and optionally subgroups and strata. groups_lists optionally specifies groupings for subgroups variables.

**Usage**

```r
extract_survival_subgroups(
  variables,
  data,
  groups_lists = list(),
  control = control_coxph(),
  label_all = "All Patients"
)
```
Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>variables</td>
<td>named list of string, list of additional analysis variables.</td>
</tr>
<tr>
<td>data</td>
<td>data.frame, the dataset containing the variables to summarize.</td>
</tr>
<tr>
<td>groups_lists</td>
<td>named list of list, optionally contains for each subgroups variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.</td>
</tr>
<tr>
<td>control</td>
<td>list, parameters for comparison details, specified by using the helper function control_coxph(). Some possible parameter options are:</td>
</tr>
<tr>
<td></td>
<td>• pval_method (string), p-value method for testing hazard ratio = 1. Default method is &quot;log-rank&quot; which comes from survival::survdiff(), can also be set to &quot;wald&quot; or &quot;likelihood&quot; (from survival::coxph()).</td>
</tr>
<tr>
<td></td>
<td>• ties (string), specifying the method for tie handling. Default is &quot;efron&quot;, can also be set to &quot;breslow&quot; or &quot;exact&quot;. See more in survival::coxph()</td>
</tr>
<tr>
<td></td>
<td>• conf_level (proportion), confidence level of the interval for HR.</td>
</tr>
<tr>
<td>label_all</td>
<td>string, label for the total population analysis.</td>
</tr>
</tbody>
</table>

Value

A named list of two elements:

• survtime: A data.frame containing columns arm, n, n_events, median, subgroup, var, var_label, and row_type.

• hr: A data.frame containing columns arm, n_tot, n_tot_events, hr, lcl, ucl, conf_level, pval, pval_label, subgroup, var, var_label, and row_type.

See Also

survival_duration_subgroups

Description

[Stable]

rtables formatting functions that handle extreme values.
Usage

h_get_format_threshold(digits = 2L)

h_format_threshold(x, digits = 2L)

Arguments

digits (integer(1))
number of decimal places to display.

x (numeric(1))
value to format.

Details

For each input, apply a format to the specified number of digits. If the value is below a threshold, it returns "<0.01" e.g. if the number of digits is 2. If the value is above a threshold, it returns ">999.99" e.g. if the number of digits is 2. If it is zero, then returns "0.00".

Value

• h_get_format_threshold() returns a list of 2 elements: threshold, with low and high thresholds, and format_string, with thresholds formatted as strings.

• h_format_threshold() returns the given value, or if the value is not within the digit threshold the relation of the given value to the digit threshold, as a formatted string.

Functions

• h_get_format_threshold(): Internal helper function to calculate the threshold and create formatted strings used in Formatting Functions. Returns a list with elements threshold and format_string.

• h_format_threshold(): Internal helper function to apply a threshold format to a value. Creates a formatted string to be used in Formatting Functions.

See Also

Other formatting functions: format_auto(), format_count_fraction(), format_count_fraction_fixed_dp(), format_count_fraction_lt10(), format_extreme_values(), format_extreme_values_ci(), format_fraction(), format_fraction_fixed_dp(), format_fraction_threshold(), format_sigfig(), format_xx(), formatting_functions

Examples

h_get_format_threshold(2L)

h_format_threshold(0.001)

h_format_threshold(1000)
**Description**

Simulated CDISC data for examples

**Usage**

- tern_ex_adsl
- tern_ex_adae
- tern_ex_adlb
- tern_ex_adpp
- tern_ex_adrs
- tern_ex_adtte

**Format**

rds(data.frame)

An object of class tbl_df (inherits from tbl, data.frame) with 200 rows and 21 columns.
An object of class tbl_df (inherits from tbl, data.frame) with 541 rows and 42 columns.
An object of class tbl_df (inherits from tbl, data.frame) with 4200 rows and 50 columns.
An object of class tbl_df (inherits from tbl, data.frame) with 522 rows and 25 columns.
An object of class tbl_df (inherits from tbl, data.frame) with 1600 rows and 29 columns.
An object of class tbl_df (inherits from tbl, data.frame) with 1000 rows and 28 columns.

**Functions**

- tern_ex_adsl: ADSL data
- tern_ex_adae: ADAE data
- tern_ex_adlb: ADLB data
- tern_ex_adpp: ADPP data
- tern_ex_adrs: ADRS data
- tern_ex_adtte: ADTTE data
Description

[Stable]

This collapses levels and only keeps those new group levels, in the order provided. The returned
factor has levels in the order given, with the possible missing level last (this will only be included if
there are missing values).

Usage

fct_collapse_only(.f, ..., .na_level = "<Missing>")

Arguments

.f (factor or character)
original vector.

... (named character)
levels in each vector provided will be collapsed into the new level given by the
respective name.

.na_level (string)
which level to use for other levels, which should be missing in the new factor.
Note that this level must not be contained in the new levels specified in . . .

Value

A modified factor with collapsed levels. Values and levels which are not included in the given
character vector input will be set to the missing level .na_level.

Note

Any existing NAs in the input vector will not be replaced by the missing level. If needed, explicit_na()
can be called separately on the result.

See Also

forcats::fctCollapse(), forcats::fct_relevel() which are used internally.

Examples

fct_collapse_only(factor(c("a", "b", "c", "d")), TRT = "b", CTRL = c("c", "d"))
**fct_discard**

_Discard specified levels of a factor_

**Description**

**[Stable]**

This discards the observations as well as the levels specified from a factor.

**Usage**

```r
fct_discard(x, discard)
```

**Arguments**

- `x` (factor)
  - the original factor.
- `discard` (character)
  - levels to discard.

**Value**

A modified factor with observations as well as levels from `discard` dropped.

**Examples**

```r
fct_discard(factor(c("a", "b", "c")), "c")
```

---

**fct_explicit_na_if**

_Insertion of explicit missing values in a factor_

**Description**

**[Stable]**

This inserts explicit missing values in a factor based on a condition. Additionally, existing NA values will be explicitly converted to given `na_level`.

**Usage**

```r
fct_explicit_na_if(x, condition, na_level = "<Missing>")
```
Arguments

x (factor)
the original factor.

condition (logical)
positions at which to insert missing values.

na_level (string)
which level to use for missing values.

Value

A modified factor with inserted and existing NA converted to na_level.

See Also

forcats::fct_na_value_to_level() which is used internally.

Examples

fct_explicit_na_if(factor(c("a", "b", NA)), c(TRUE, FALSE, FALSE))

Description

[Stable]
Fitting functions for univariate and multivariate Cox regression models.

Usage

fit_coxreg_univar(variables, data, at = list(), control = control_coxreg())

fit_coxreg_multivar(variables, data, control = control_coxreg())

Arguments

variables (named list)
the names of the variables found in data, passed as a named list and corresponding to the time, event, arm, strata, and covariates terms. If arm is missing from variables, then only Cox model(s) including the covariates will be fitted and the corresponding effect estimates will be tabulated later.

data (data.frame)
the dataset containing the variables to fit the models.
at (list of numeric)
when the candidate covariate is a numeric, use at to specify the value of the
covariate at which the effect should be estimated.

control (list)
a list of parameters as returned by the helper function control_coxreg().

Value
• fit_coxreg_univar() returns a coxreg.univar class object which is a named list with 5 elements:
  – mod: Cox regression models fitted by survival::coxph().
  – data: The original data frame input.
  – control: The original control input.
  – vars: The variables used in the model.
  – at: Value of the covariate at which the effect should be estimated.

• fit_coxreg_multivar() returns a coxreg.multivar class object which is a named list with 4 elements:
  – mod: Cox regression model fitted by survival::coxph().
  – data: The original data frame input.
  – control: The original control input.
  – vars: The variables used in the model.

Functions
• fit_coxreg_univar(): Fit a series of univariate Cox regression models given the inputs.
• fit_coxreg_multivar(): Fit a multivariate Cox regression model.

Note
When using fit_coxreg_univar there should be two study arms.

See Also
h_cox_regression for relevant helper functions, cox_regression.

Examples
library(survival)
set.seed(1, kind = "Mersenne-Twister")

# Testing dataset [survival::bladder].
dta_bladder <- with(
data = bladder[bladder$enum < 5, ],
data.frame(
time = stop,
status = event,
armcd = as.factor(rx),
covar1 = as.factor(enum),
covar2 = factor(
    sample(as.factor(enum)),
    levels = 1:4, labels = c("F", "F", "M", "M")
)
)
)
labels <- c("armcd" = "ARM", "covar1" = "A Covariate Label", "covar2" = "Sex (F/M)"
formatters::var_labels(dta_bladder)[names(labels)] <- labels
dta_bladder$age <- sample(20:60, size = nrow(dta_bladder), replace = TRUE)

plot(
    survfit(Surv(time, status) ~ armcd + covar1, data = dta_bladder),
    lty = 2:4,
    xlab = "Months",
    col = c("blue1", "blue2", "blue3", "blue4", "red1", "red2", "red3", "red4")
)

# fit_coxreg_univar

## Cox regression: arm + 1 covariate.
mod1 <- fit_coxreg_univar(
    variables = list(
        time = "time", event = "status", arm = "armcd",
        covariates = "covar1"
    ),
    data = dta_bladder,
    control = control_coxreg(conf_level = 0.91)
)

## Cox regression: arm + 1 covariate + interaction, 2 candidate covariates.
mod2 <- fit_coxreg_univar(
    variables = list(
        time = "time", event = "status", arm = "armcd",
        covariates = c("covar1", "covar2"
    ),
    data = dta_bladder,
    control = control_coxreg(conf_level = 0.91, interaction = TRUE)
)

## Cox regression: arm + 1 covariate, stratified analysis.
mod3 <- fit_coxreg_univar(
    variables = list(
        time = "time", event = "status", arm = "armcd", strata = "covar2",
        covariates = c("covar1"
    ),
    data = dta_bladder,
    control = control_coxreg(conf_level = 0.91)
)

## Cox regression: no arm, only covariates.
mod4 <- fit_coxreg_univar(
variables = list(
    time = "time", event = "status",
    covariates = c("covar1", "covar2")
),
data = dta_bladder)

# fit_coxreg_multivar

## Cox regression: multivariate Cox regression.
multivar_model <- fit_coxreg_multivar(
    variables = list(
        time = "time", event = "status", arm = "armcd",
        covariates = c("covar1", "covar2")
    ),
data = dta_bladder)

# Example without treatment arm.
multivar_covs_model <- fit_coxreg_multivar(
    variables = list(
        time = "time", event = "status",
        covariates = c("covar1", "covar2")
    ),
data = dta_bladder)

---

**fit_logistic**

*Fit for logistic regression*

**Description**

[Stable]

Fit a (conditional) logistic regression model.

**Usage**

```r
fit_logistic( 
data, 
variables = list(response = "Response", arm = "ARMCD", covariates = NULL, interaction = NULL, strata = NULL), 
response_definition = "response"
)
```

**Arguments**

- `data` (data.frame)
  
  the data frame on which the model was fit.
variables (named list of string)
  list of additional analysis variables.

response_definition (string)
  the definition of what an event is in terms of response. This will be used when
  fitting the (conditional) logistic regression model on the left hand side of the
  formula.

Value
  A fitted logistic regression model.

Model Specification
  The variables list needs to include the following elements:

  • arm: Treatment arm variable name.
  • response: The response arm variable name. Usually this is a 0/1 variable.
  • covariates: This is either NULL (no covariates) or a character vector of covariate variable
    names.
  • interaction: This is either NULL (no interaction) or a string of a single covariate variable
    name already included in covariates. Then the interaction with the treatment arm is included
    in the model.

Examples

library(dplyr)

adrs_f <- tern_ex_adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(RACE %in% c("ASIAN", "WHITE", "BLACK OR AFRICAN AMERICAN")) %>%
  mutate(
    Response = case_when(AVALC %in% c("PR", "CR") ~ 1, TRUE ~ 0),
    RACE = factor(RACE),
    SEX = factor(SEX)
  )
formatters::var_labels(adrs_f) <- c(formatters::var_labels(tern_ex_adrs), Response = "Response")
mod1 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE")
  )
)
mod2 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    interaction = "RACE"
  )
)
fit_rsp_step

covariates = c("AGE", "RACE"),
interaction = "AGE"
)
)

Description

[Stable]
This fits the Subgroup Treatment Effect Pattern logistic regression models for a binary (response) outcome. The treatment arm variable must have exactly 2 levels, where the first one is taken as reference and the estimated odds ratios are for the comparison of the second level vs. the first one.

The (conditional) logistic regression model which is fit is:
response ~ arm * poly(biomarker, degree) + covariates + strata(strata)
where degree is specified by control_step().

Usage

fit_rsp_step(variables, data, control = c(control_step(), control_logistic()))

Arguments

variables (named list of character)
list of analysis variables: needs response, arm, biomarker, and optional covariates and strata.
data (data.frame)
the dataset containing the variables to summarize.
control (named list)
combined control list from control_step() and control_logistic().

Value

A matrix of class step. The first part of the columns describe the subgroup intervals used for the biomarker variable, including where the center of the intervals are and their bounds. The second part of the columns contain the estimates for the treatment arm comparison.

Note

For the default degree 0 the biomarker variable is not included in the model.

See Also

control_step() and control_logistic() for the available customization options.
Examples

# Testing dataset with just two treatment arms.
library(survival)
library(dplyr)

adrs_f <- tern_ex_adrs %>%
  filter(
    PARAMCD == "BESRSP1",
    ARM %in% c("B: Placebo", "A: Drug X")
  ) %>%
  mutate(
    # Reorder levels of ARM to have Placebo as reference arm for Odds Ratio calculations.
    ARM = droplevels(forcats::fct_relevel(ARM, "B: Placebo")),
    RSP = case_when(AVALC %in% c("PR", "CR") ~ 1, TRUE ~ 0),
    SEX = factor(SEX)
  )

variables <- list(
  arm = "ARM",
  biomarker = "BMRKR1",
  covariates = "AGE",
  response = "RSP"
)

# Fit default STEP models: Here a constant treatment effect is estimated in each subgroup.
# We use a large enough bandwidth to avoid too small subgroups and linear separation in those.
step_matrix <- fit_rsp_step(
  variables = variables,
  data = adrs_f,
  control = c(control_logistic(), control_step(bandwidth = 0.9))
)
dim(step_matrix)
head(step_matrix)

# Specify different polynomial degree for the biomarker interaction to use more flexible local
# models. Or specify different logistic regression options, including confidence level.
step_matrix2 <- fit_rsp_step(
  variables = variables,
  data = adrs_f,
  control = c(control_logistic(conf_level = 0.9), control_step(bandwidth = NULL, degree = 1))
)

# Use a global constant model. This is helpful as a reference for the subgroup models.
step_matrix3 <- fit_rsp_step(
  variables = variables,
  data = adrs_f,
  control = c(control_logistic(), control_step(bandwidth = NULL, num_points = 2L))
)

# It is also possible to use strata, i.e. use conditional logistic regression models.
variables2 <- list(
  arm = "ARM",

fit_survival_step

```
b biomarker = "BMRKR1",
 covariates = "AGE",
 response = "RSP",
 strata = c("STRATA1", "STRATA2")
```

```
step_matrix4 <- fit_rsp_step(
  variables = variables2,
  data = adrs_f,
  control = c(control_logistic(), control_step(bandwidth = NULL))
)
```

---

fit_survival_step  Subgroup treatment effect pattern (STEP) fit for survival outcome

**Description**

[Stable]

This fits the subgroup treatment effect pattern (STEP) models for a survival outcome. The treatment arm variable must have exactly 2 levels, where the first one is taken as reference and the estimated hazard ratios are for the comparison of the second level vs. the first one.

The model which is fit is:

```
Surv(time, event) ~ arm * poly(biomarker, degree) + covariates + strata(strata)
```

where degree is specified by `control_step()`.

**Usage**

```
fit_survival_step(
  variables,
  data,
  control = c(control_step(), control_coxph())
)
```

**Arguments**

- **variables** *(named list of character)*
  - list of analysis variables: needs time, event, arm, biomarker, and optional covariates and strata.

- **data** *(data.frame)*
  - the dataset containing the variables to summarize.

- **control** *(named list)*
  - combined control list from `control_step()` and `control_coxph()`.
Value

A matrix of class step. The first part of the columns describe the subgroup intervals used for the biomarker variable, including where the center of the intervals are and their bounds. The second part of the columns contain the estimates for the treatment arm comparison.

Note

For the default degree 0 the biomarker variable is not included in the model.

See Also

control_step() and control_coxph() for the available customization options.

Examples

# Testing dataset with just two treatment arms.
library(dplyr)

adtte_f <- tern_ex_adtte %>%
  filter(
    PARAMCD == "OS",
    ARM %in% c("B: Placebo", "A: Drug X")
  ) %>%
  mutate(
    # Reorder levels of ARM to display reference arm before treatment arm.
    ARM = droplevels(forcats::fct_relevel(ARM, "B: Placebo")),
    is_event = CNSR == 0
  )
layers <- c("ARM" = "Treatment Arm", "is_event" = "Event Flag")
formatters::var_labels(adtte_f)[names(lables)] <- layers

variables <- list(
  arm = "ARM",
  biomarker = "BMRKR1",
  covariates = c("AGE", "BMRKR2"),
  event = "is_event",
  time = "AVAL"
)

# Fit default STEP models: Here a constant treatment effect is estimated in each subgroup.
step_matrix <- fit_survival_step(
  variables = variables,
  data = adtte_f
)
dim(step_matrix)
head(step_matrix)

# Specify different polynomial degree for the biomarker interaction to use more flexible local
# models. Or specify different Cox regression options.
step_matrix2 <- fit_survival_step(
  variables = variables,
  data = adtte_f,
control = c(control_coxph(conf_level = 0.9), control_step(degree = 2))

# Use a global model with cubic interaction and only 5 points.
step_matrix3 <- fit_survival_step(
  variables = variables,
  data = adtte_f,
  control = c(control_coxph(), control_step(bandwidth = NULL, degree = 3, num_points = 5L))
)

forest_viewport

Create a viewport tree for the forest plot

Description
[Deprecated]

Usage

forest_viewport(
  tbl,
  width_row_names = NULL,
  width_columns = NULL,
  width_forest = grid::unit(1, "null"),
  gap_column = grid::unit(1, "lines"),
  gap_header = grid::unit(1, "lines"),
  mat_form = NULL
)

Arguments

tbl (VTableTree)
tables table object.
width_row_names (grid::unit)
width of row names.
width_columns (grid::unit)
width of column spans.
width_forest (grid::unit)
width of the forest plot.
gap_column (grid::unit)
gap width between the columns.
gap_header (grid::unit)
gap width between the header.
mat_form (MatrixPrintForm)
matrix print form of the table.
Value

A viewport tree.

Examples

```r
library(grid)

tbl <- rtable(
  header = rheader(
    rrow("", "E", rcell("CI", colspan = 2)),
    rrow("", "A", "B", "C")
  ),
  rrow("row 1", 1, 0.8, 1.1),
  rrow("row 2", 1.4, 0.8, 1.6),
  rrow("row 3", 1.2, 0.8, 1.2)
)

v <- forest_viewport(tbl)

grid::grid.newpage()
showViewport(v)
```

formatting_functions  Formatting functions

Description

[Stable]

See below for the list of formatting functions created in tern to work with rtables.

Other available formats can be listed via `formatters::list_valid_format_labels()`. Additional custom formats can be created via the `formatters::sprintf_format()` function.

See Also

Other formatting functions: `extreme_format`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lt10()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`, `format_fraction_threshold()`, `format_sigfig()`, `format_xx()`
**format_auto**

Format automatically using data significant digits

---

**Description**

**[Stable]**

Formatting function for the majority of default methods used in `analyze_vars()`. For non-derived values, the significant digits of data is used (e.g. range), while derived values have one more digits (measure of location and dispersion like mean, standard deviation). This function can be called internally with "auto" like, for example, `.formats = c("mean" = "auto")`. See details to see how this works with the inner function.

**Usage**

```r
format_auto(dt_var, x_stat)
```

**Arguments**

- `dt_var` (numeric)
  variable data the statistics were calculated from. Used only to find significant digits. In `analyze_vars` this comes from `.df_row` (see `rtables::additional_fun_params`), and it is the row data after the above row splits. No column split is considered.

- `x_stat` (string)
  string indicating the current statistical method used.

**Details**

The internal function is needed to work with `rtables` default structure for format functions, i.e. function(x, ...), where is x are results from statistical evaluation. It can be more than one element (e.g. for `.stats = "mean_sd"`).

**Value**

A string that `rtables` prints in a table cell.

**See Also**

Other formatting functions: `extreme_format`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lt10()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`, `format_fraction_threshold()`, `format_sigfig()`, `format_xx()`, `formatting_functions`
format_count_fraction  Format count and fraction

Description

[Stable]

Formats a count together with fraction with special consideration when count is 0.

Usage

format_count_fraction(x, ...)

Arguments

x (numeric(2))

vector of length 2 with count and fraction, respectively.

... not used. Required for rtables interface.

Value

A string in the format count (fraction %). If count is 0, the format is 0.

See Also

Other formatting functions: extreme_format, format_auto(), format_count_fraction_fixed_dp(), format_count_fraction_lt10(), format_extreme_values(), format_extreme_values_ci(), format_fraction(), format_fraction_fixed_dp(), format_fraction_threshold(), format_sigfig(), format_xx(), formatting_functions

Examples

format_count_fraction(x = c(2, 0.6667))
format_count_fraction(x = c(0, 0))
**format_count_fraction_fixed_dp**

*Format count and percentage with fixed single decimal place*

**Description**

[Experimental]
Formats a count together with fraction with special consideration when count is 0.

**Usage**

```r
format_count_fraction_fixed_dp(x, ...)
```

**Arguments**

```
x numeric(2) vector of length 2 with count and fraction, respectively.
... not used. Required for rtables interface.
```

**Value**

A string in the format count (fraction %). If count is 0, the format is 0.

**See Also**

Other formatting functions:
- `extreme_format`
- `format_auto`
- `format_count_fraction`
- `format_count_fraction_lt10`
- `format_extreme_values`
- `format_extreme_values_ci`
- `format_fraction`
- `format_fraction_fixed_dp`
- `format_fraction_threshold`
- `format_sigfig`
- `format_xx`
- `formatting_functions`

**Examples**

```r
format_count_fraction_fixed_dp(x = c(2, 0.6667))
format_count_fraction_fixed_dp(x = c(2, 0.5))
format_count_fraction_fixed_dp(x = c(0, 0))
```

---

**format_count_fraction_lt10**

*Format count and fraction with special case for count < 10*

**Description**

[Stable]
Formats a count together with fraction with special consideration when count is less than 10.
Usage

format_count_fraction_lt10(x, ...)

Arguments

x (numeric(2))
vector of length 2 with count and fraction, respectively.
...
not used. Required for rtables interface.

Value

A string in the format count (fraction %). If count is less than 10, only count is printed.

See Also

Other formatting functions: extreme_format, format_auto(), format_count_fraction(), format_count_fraction_fixed_dp(), format_extreme_values(), format_extreme_values_ci(), format_fraction(), format_fraction_fixed_dp(), format_fraction_threshold(), format_sigfig(), format_xx(), formatting_functions

Examples

format_count_fraction_lt10(x = c(275, 0.9673))
format_count_fraction_lt10(x = c(2, 0.6667))
format_count_fraction_lt10(x = c(9, 1))
See Also

Other formatting functions: `format_extreme_values`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`,
`format_count_fraction_lt10()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`,
`format_fraction_threshold()`, `format_sigfig()`, `format_xx()`, `formatting_functions`

Examples

```r
format_fun <- format_extreme_values(2L)
format_fun(x = 0.127)
format_fun(x = Inf)
format_fun(x = 0)
format_fun(x = 0.009)
```

---

`format_extreme_values_ci`

*Format extreme values part of a confidence interval*

Description

[Stable]

Formatting Function for extreme values part of a confidence interval. Values are formatted as e.g.
"(xx.xx, xx.xx)" if the number of digits is 2.

Usage

```
format_extreme_values_ci(digits = 2L)
```

Arguments

- **digits** (`integer(1)`)  number of decimal places to display.

Value

An `rtables` formatting function that uses threshold digits to return a formatted extreme values
confidence interval.

See Also

Other formatting functions: `format_extreme_values`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`,
`format_count_fraction_lt10()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`,
`format_fraction_threshold()`, `format_sigfig()`, `format_xx()`, `formatting_functions`
format_fraction

Examples

```r
format_fun <- format_extreme_values_ci(2L)
format_fun(x = c(0.127, Inf))
format_fun(x = c(0, 0.009))
```

---

**Description**

[Stable]

Formats a fraction together with ratio in percent.

**Usage**

```r
format_fraction(x, ...)
```

**Arguments**

- `x` (named integer)
  - vector with elements `num` and `denom`.
- `...` not used. Required for `rtables` interface.

**Value**

A string in the format `num / denom (ratio %)`. If `num` is 0, the format is `num / denom`.

**See Also**

Other formatting functions: `extreme_format`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lt10()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction_fixed_dp()`, `format_fraction_threshold()`, `format_sigfig()`, `format_xx()`, `formatting_functions`

**Examples**

```r
format_fraction(x = c(num = 2L, denom = 3L))
format_fraction(x = c(num = 0L, denom = 3L))
```
**Description**

[Stable]

Formats a fraction together with ratio in percent with fixed single decimal place. Includes trailing zero in case of whole number percentages to always keep one decimal place.

**Usage**

```r
format_fraction_fixed_dp(x, ...)
```

**Arguments**

- `x` (named integer)
  - vector with elements `num` and `denom`.
- `...` not used. Required for `rtables` interface.

**Value**

A string in the format `num / denom (ratio %)`. If `num` is 0, the format is `num / denom`.

**See Also**

Other formatting functions: `extreme_format`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lt10()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_threshold()`, `format_sigfig()`, `format_xx()`, `formatting_functions`

**Examples**

```r
format_fraction_fixed_dp(x = c(num = 1L, denom = 2L))
format_fraction_fixed_dp(x = c(num = 1L, denom = 4L))
format_fraction_fixed_dp(x = c(num = 0L, denom = 3L))
```
**format_fraction_threshold**

*Format fraction with lower threshold*

**Description**

[Stable]

Formats a fraction when the second element of the input `x` is the fraction. It applies a lower threshold, below which it is just stated that the fraction is smaller than that.

**Usage**

```r
format_fraction_threshold(threshold)
```

**Arguments**

- `threshold`: (proportion)
  - lower threshold.

**Value**

An `rtables` formatting function that takes numeric input `x` where the second element is the fraction that is formatted. If the fraction is above or equal to the threshold, then it is displayed in percentage. If it is positive but below the threshold, it returns, e.g. "<1" if the threshold is 0.01. If it is zero, then just "0" is returned.

**See Also**

Other formatting functions: `extreme_format`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lt10()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`, `format_sigfig()`, `format_xx()`, `formatting_functions`

**Examples**

```r
format_fun <- format_fraction_threshold(0.05)
format_fun(x = c(20, 0.1))
format_fun(x = c(2, 0.01))
format_fun(x = c(0, 0))
```
format_sigfig

Format numeric values by significant figures

Description

Format numeric values to print with a specified number of significant figures.

Usage

format_sigfig(sigfig, format = "xx", num_fmt = "fg")

Arguments

- **sigfig** (integer(1))
  number of significant figures to display.

- **format** (string)
  the format label (string) to apply when printing the value. Decimal places in string are ignored in favor of formatting by significant figures. Formats options are: "xx", "xx / xx", "(xx, xx)", "xx - xx", and "xx (xx)".

- **num_fmt** (string)
  numeric format modifiers to apply to the value. Defaults to "fg" for standard significant figures formatting - fixed (non-scientific notation) format ("f") and sigfig equal to number of significant figures instead of decimal places ("g"). See the formatC() format argument for more options.

Value

An rtables formatting function.

See Also

Other formatting functions: extreme_format, format_auto(), format_count_fraction(), format_count_fraction_fixed_dp(), format_count_fraction_lt10(), format_extreme_values(), format_extreme_values_ci(), format_fraction(), format_fraction_fixed_dp(), format_fraction_threshold(), format_xx(), formatting_functions

Examples

fmt_3sf <- format_sigfig(3)
fmt_3sf(1.658)
fmt_3sf(1e1)

fmt_5sf <- format_sigfig(5)
fmt_5sf(0.57)
fmt_5sf(0.000025645)
Format XX as a formatting function

Description

Translate a string where x and dots are interpreted as number place holders, and others as formatting elements.

Usage

```r
format_xx(str)
```

Arguments

- `str` (string) template.

Value

An `rtables` formatting function.

See Also

Other formatting functions: `extreme_format`, `format_auto()`, `format_count_fraction()`, `format_count_fraction_fixed_dp()`, `format_count_fraction_lti0()`, `format_extreme_values()`, `format_extreme_values_ci()`, `format_fraction()`, `format_fraction_fixed_dp()`, `format_fraction_threshold()`, `format_sigfig()`, `formatting_functions`

Examples

```r
test <- list(c(1.658, 0.5761), c(1e1, 785.6))

z <- format_xx("xx (xx.x)")
sapply(test, z)

z <- format_xx("xx.x - xx.x")
sapply(test, z)

z <- format_xx("xx.x, incl. xx.x% NE")
sapply(test, z)
```
f_conf_level  
Utility function to create label for confidence interval

Description

[Stable]

Usage

f_conf_level(conf_level)

Arguments

conf_level  
(proportion)
confidence level of the interval.

Value

A string.

f_pval  
Utility function to create label for p-value

Description

[Stable]

Usage

f_pval(test_mean)

Arguments

test_mean  
(numeric(1))
mean value to test under the null hypothesis.

Value

A string.
### get_smooths

**Description**

[Stable]

This produces loess smoothed estimates of y with Student confidence intervals.

**Usage**

```r
get_smooths(df, x, y, groups = NULL, level = 0.95)
```

**Arguments**

- `df` *(data.frame)*
  - data set containing all analysis variables.
- `x` *(string)*
  - x column name.
- `y` *(string)*
  - y column name.
- `groups` *(character or NULL)*
  - vector with optional grouping variables names.
- `level` *(proportion)*
  - level of confidence interval to use (0.95 by default).

**Value**

A `data.frame` with original x, smoothed y, ylow, and yhigh, and optional groups variables formatted as factor type.

### groups_list_to_df

**Description**

This converts a list of group levels into a data frame format which is expected by `rtables::add_combo_levels()`.

**Usage**

```r
groups_list_to_df(groups_list)
```

**Arguments**

- `groups_list` *(named list of character)*
  - specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.
**Value**

A tibble in the required format.

**Examples**

```r
grade_groups <- list(
  "Any Grade (%)" = c("1", "2", "3", "4", "5"),
  "Grade 3-4 (%)" = c("3", "4"),
  "Grade 5 (%)" = "5"
)
groups_list_to_df(grade_groups)
```

---

**Description**

[Stable]

**Usage**

```r
g_forest(
  tbl, 
  col_x = attr(tbl, "col_x"),
  col_ci = attr(tbl, "col_ci"),
  vline = 1,
  forest_header = attr(tbl, "forest_header"),
  xlim = c(0.1, 10),
  logx = TRUE,
  x_at = c(0.1, 1, 10),
  width_row_names = lifecycle::deprecated(),
  width_columns = NULL,
  width_forest = lifecycle::deprecated(),
  lbl_col_padding = 0,
  rel_width_forest = 0.25,
  font_size = 12,
  col_symbol_size = attr(tbl, "col_symbol_size"),
  col = getOption("ggplot2.discrete.colour")[1],
  ggtheme = NULL,
  as_list = FALSE,
  gp = lifecycle::deprecated(),
  draw = lifecycle::deprecated(),
  newpage = lifecycle::deprecated()
)
```
Arguments

`tbl` (VTableTree)
- `rtables` table with at least one column with a single value and one column with 2 values.

`col_x` (integer(1) or NULL)
- Column index with estimator. By default tries to get this from `tbl` attribute `col_x`, otherwise needs to be manually specified. If NULL, points will be excluded from forest plot.

`col_ci` (integer(1) or NULL)
- Column index with confidence intervals. By default tries to get this from `tbl` attribute `col_ci`, otherwise needs to be manually specified. If NULL, lines will be excluded from forest plot.

`vline` (numeric(1) or NULL)
- X coordinate for vertical line, if NULL then the line is omitted.

`forest_header` (character(2))
- Text displayed to the left and right of `vline`, respectively. If `vline` = NULL then `forest_header` is not printed. By default tries to get this from `tbl` attribute `forest_header`. If NULL, defaults will be extracted from the table if possible, and set to "Comparison\nBetter" and "Treatment\nBetter" if not.

`xlim` (numeric(2))
- Limits for x axis.

`logx` (flag)
- Show the x-values on logarithm scale.

`x_at` (numeric)
- X-tick locations, if NULL, `x_at` is set to `vline` and both `xlim` values.

`width_row_names`
- [Deprecated] Please use the `lbl_col_padding` argument instead.

`width_columns` (numeric)
- A vector of column widths. Each element's position in `colwidths` corresponds to the column of `tbl` in the same position. If NULL, column widths are calculated according to maximum number of characters per column.

`width_forest`
- [Deprecated] Please use the `rel_width_forest` argument instead.

`lbl_col_padding` (numeric)
- Additional padding to use when calculating spacing between the first (label) column and the second column of `tbl`. If `colwidths` is specified, the width of the first column becomes `colwidths[1] + lbl_col_padding`. Defaults to 0.

`rel_width_forest` (proportion)
- Proportion of total width to allocate to the forest plot. Relative width of table is then 1 - `rel_width_forest`. If `as_list = TRUE`, this parameter is ignored.

`font_size` (numeric(1))
- Font size.
col_symbol_size
  (numeric or NULL)
  column index from tbl containing data to be used to determine relative size for
estimator plot symbol. Typically, the symbol size is proportional to the sample
size used to calculate the estimator. If NULL, the same symbol size is used for
all subgroups. By default tries to get this from tbl attribute col_symbol_size,
otherwise needs to be manually specified.

col
  (character)
  color(s).

ggtheme
  (theme)
  a graphical theme as provided by ggplot2 to control styling of the plot.

as_list
  (flag)
  whether the two ggplot objects should be returned as a list. If TRUE, a named
list with two elements, table and plot, will be returned. If FALSE (default) the
table and forest plot are printed side-by-side via cowplot::plot_grid().

gp
  [Deprecated] g_forest is now generated as a ggplot object. This argument is
  no longer used.

draw
  [Deprecated] g_forest is now generated as a ggplot object. This argument is
  no longer used.

newpage
  [Deprecated] g_forest is now generated as a ggplot object. This argument is
  no longer used.

Details

Given a rtables::rtable() object with at least one column with a single value and one column
with 2 values, converts table to a ggplot2::ggplot() object and generates an accompanying forest
plot. The table and forest plot are printed side-by-side.

Value

ggplot forest plot and table.

Examples

library(dplyr)
library(forcats)
library(nestcolor)

adrs <- tern_ex_adrs
n_records <- 20
adrs_labels <- formatters::var_labels(adrs, fill = TRUE)
adrs <- adrs %>%
  filter(PARAMCD == "BESRSP1") %>%
  filter(ARM %in% c("A: Drug X", "B: Placebo")) %>%
  slice(seq_len(n_records)) %>%
  droplevels() %>%
  mutate(
    # Reorder levels of factor to make the placebo group the reference arm.
    ARM = fct_relevel(ARM, "B: Placebo"),

    col_symbol_size
    col
    ggtheme
    as_list
    gp
    draw
    newpage
rsp = AVALC == "CR"
)
formatters::var_labels(adrs) <- c(adrs_labels, "Response")
df <- extract_rsp_subgroups(
  variables = list(rsp = "rsp", arm = "ARM", subgroups = c("SEX", "STRATA2")),
  data = adrs
)
# Full commonly used response table.

tbl <- basic_table() %>%
  tabulate_rsp_subgroups(df)
g_forest(tbl)

# Odds ratio only table.

tbl_or <- basic_table() %>%
  tabulate_rsp_subgroups(df, vars = c("n_tot", "or", "ci"))
g_forest(
  tbl_or,
  forest_header = c("Comparison
Better", "Treatment
Better")
)

# Survival forest plot example.
adtte <- tern_ex_adtte

# Save variable labels before data processing steps.
adtte_labels <- formatters::var_labels(adtte, fill = TRUE)
adtte_f <- adtte %>%
  filter(
    PARAMCD == "OS",
    ARM %in% c("B: Placebo", "A: Drug X"),
    SEX %in% c("M", "F")
  )
  # Reorder levels of ARM to display reference arm before treatment arm.
  ARM = droplevels(fct_relevel(ARM, "B: Placebo")),
  SEX = droplevels(SEX),
  AVALU = as.character(AVALU),
  is_event = CNSR == 0
)
labels <- list(
  "ARM" = adtte_labels["ARM"],
  "SEX" = adtte_labels["SEX"],
  "AVALU" = adtte_labels["AVALU"],
  "is_event" = "Event Flag"
)
formatters::var_labels(adtte_f)[names(labels)] <- as.character(labels)
df <- extract_survival_subgroups(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    arm = "ARM", subgroups = c("SEX", "BMRKR2")
  ),
  data = adtte_f
Individual patient plots

Description

[Stable]

Line plot(s) displaying trend in patients’ parameter values over time is rendered. Patients’ individual baseline values can be added to the plot(s) as reference.
Usage

g_ipp(
  df,
  xvar,
  yvar,
  xlab,
  ylab,
  id_var = "USUBJID",
  title = "Individual Patient Plots",
  subtitle = "",
  caption = NULL,
  add_baseline_hline = FALSE,
  yvar_baseline = "BASE",
  ggtheme = nestcolor::theme_nest(),
  plotting_choices = c("all_in_one", "split_by_max_obs", "separate_by_obs"),
  max_obs_per_plot = 4,
  col = NULL
)

Arguments

df (data.frame)
data set containing all analysis variables.

xvar (string)
time point variable to be plotted on x-axis.

yvar (string)
continuous analysis variable to be plotted on y-axis.

xlab (string)
plot label for x-axis.

ylab (string)
plot label for y-axis.

id_var (string)
variable used as patient identifier.

title (string)
title for plot.

subtitle (string)
subtitle for plot.

caption (string)
optional caption below the plot.

add_baseline_hline (flag)
adds horizontal line at baseline y-value on plot when TRUE.

yvar_baseline (string)
variable with baseline values only. Ignored when addBaseline_hline is FALSE.
g_ipp

**ggtheme** (theme)
optional graphical theme function as provided by ggplot2 to control outlook of plot. Use ggplot2::theme() to tweak the display.

**plotting_choices** (string)
specifies options for displaying plots. Must be one of "all_in_one", "split_by_max_obs", or "separate_by_obs".

**max_obs_per_plot** (integer(1))
number of observations to be plotted on one plot. Ignored if plotting_choices is not "separate_by_obs".

**col** (character)
line colors.

**Value**
A ggplot object or a list of ggplot objects.

**Functions**
- **g_ipp()**: Plotting function for individual patient plots which, depending on user preference, renders a single graphic or compiles a list of graphics that show trends in individual’s parameter values over time.

**See Also**
Relevant helper function h_g_ipp().

**Examples**

```r
library(dplyr)
library(nestcolor)

# Select a small sample of data to plot.
adlb <- tern_ex_adlb %>%
  filter(PARAMCD == "ALT", !(AVISIT %in% c("SCREENING", "BASELINE"))) %>%
slice(1:36)

plot_list <- g_ipp(
df = adlb,
xvar = "AVISIT",
yvar = "AVAL",
xlab = "Visit",
ylab = "SGOT/ALT (U/L)",
title = "Individual Patient Plots",
add_baseline_hline = TRUE,
plotting_choices = "split_by_max_obs",
max_obs_per_plot = 5
)
plot_list
```
Description

[Stable]
From a survival model, a graphic is rendered along with tabulated annotation including the number of patient at risk at given time and the median survival per group.

Usage

```R
g_km(
  df,
  variables,
  control_surv = control_surv_timepoint(),
  col = NULL,
  lty = NULL,
  lwd = 0.5,
  censor_show = TRUE,
  pch = 3,
  size = 2,
  max_time = NULL,
  xticks = NULL,
  xlab = "Days",
  yval = c("Survival", "Failure"),
  ylab = paste(yval, "Probability"),
  ylim = NULL,
  title = NULL,
  footnotes = NULL,
  font_size = 10,
  ci_ribbon = FALSE,
  annot_at_risk = TRUE,
  annot_at_risk_title = TRUE,
  annot_surv_med = TRUE,
  annot_coxph = FALSE,
  annot_stats = NULL,
  annot_stats_vlines = FALSE,
  control_coxph_pw = control_coxph(),
  ref_group_coxph = NULL,
  control_annot_surv_med = control_surv_med_annot(),
  control_annot_coxph = control_coxph_annot(),
  legend_pos = NULL,
  rel_height_plot = 0.75,
  ggtheme = NULL,
  as_list = FALSE,
  draw = lifecycle::deprecated(),
)```

newpage = lifecycle::deprecated(),
gp = lifecycle::deprecated(),
vp = lifecycle::deprecated(),
name = lifecycle::deprecated(),
annot_coxph_ref_lbls = lifecycle::deprecated(),
position_coxph = lifecycle::deprecated(),
position_surv_med = lifecycle::deprecated(),
width_annots = lifecycle::deprecated()
)

Arguments

**df** *(data.frame)*
data set containing all analysis variables.

**variables** *(named list)*
variable names. Details are:
- **tte** *(numeric)*
  variable indicating time-to-event duration values.
- **is_event** *(logical)*
  event variable. TRUE if event, FALSE if time to event is censored.
- **arm** *(factor)*
  the treatment group variable.
- **strata** *(character or NULL)*
  variable names indicating stratification factors.

**control_surv** *(list)*
parameters for comparison details, specified by using the helper function `control_surv_timepoint()`. Some possible parameter options are:
- **conf_level** *(proportion)*
  confidence level of the interval for survival rate.
- **conf_type** *(string)*
  "plain" (default), "log", "log-log" for confidence interval type, see more in `survival::survfit()`. Note that the option "none" is no longer supported.

**col** *(character)*
lines colors. Length of a vector should be equal to number of strata from `survival::survfit()`.

**lty** *(numeric)*
line type. If a vector is given, its length should be equal to the number of strata from `survival::survfit()`.

**lwd** *(numeric)*
line width. If a vector is given, its length should be equal to the number of strata from `survival::survfit()`.

**censor_show** *(flag)*
whether to show censored observations.

**pch** *(string)*
name of symbol or character to use as point symbol to indicate censored cases.
size (numeric(1))
size of censored point symbols.

max_time (numeric(1))
maximum value to show on x-axis. Only data values less than or up to this threshold value will be plotted (defaults to NULL).

xticks (numeric or NULL)
numeric vector of tick positions or a single number with spacing between ticks on the x-axis. If NULL (default), `labeling::extended()` is used to determine optimal tick positions on the x-axis.

xlab (string)
x-axis label.

yval (string)
type of plot, to be plotted on the y-axis. Options are Survival (default) and Failure probability.

ylab (string)
y-axis label.

ylim (numeric(2))
vector containing lower and upper limits for the y-axis, respectively. If NULL (default), the default scale range is used.

title (string)
plot title.

footnotes (string)
plot footnotes.

font_size (numeric(1))
font size to use for all text.

ci_ribbon (flag)
whether the confidence interval should be drawn around the Kaplan-Meier curve.

annot_at_risk (flag)
compute and add the annotation table reporting the number of patient at risk matching the main grid of the Kaplan-Meier curve.

annot_at_risk_title (flag)
whether the "Patients at Risk" title should be added above the annot_at_risk table. Has no effect if annot_at_risk is FALSE. Defaults to TRUE.

annot_surv_med (flag)
compute and add the annotation table on the Kaplan-Meier curve estimating the median survival time per group.

annot_coxph (flag)
whether to add the annotation table from a `survival::coxph()` model.

annot_stats (string or NULL)
statistics annotations to add to the plot. Options are median (median survival follow-up time) and min (minimum survival follow-up time).
annot_stats_vlines
(flag)
add vertical lines corresponding to each of the statistics specified by annot_stats. If annot_stats is NULL no lines will be added.

control_coxph_pw
(list)
parameters for comparison details, specified using the helper function control_coxph(). Some possible parameter options are:

• pval_method (string)
p-value method for testing hazard ratio = 1. Default method is "log-rank", can also be set to "wald" or "likelihood".
• ties (string)
method for tie handling. Default is "efron", can also be set to "breslow" or "exact". See more in survival::coxph()
• conf_level (proportion)
confidence level of the interval for HR.

ref_group_coxph
(string or NULL)
level of arm variable to use as reference group in calculations for annot_coxph table. If NULL (default), uses the first level of the arm variable.

control_annot_surv_med
(list)
parameters to control the position and size of the annotation table added to the plot when annot_surv_med = TRUE, specified using the control_surv_med_annot() function. Parameter options are: x, y, w, h, and fill. See control_surv_med_annot() for details.

control_annot_coxph
(list)
parameters to control the position and size of the annotation table added to the plot when annot_coxph = TRUE, specified using the control_coxph_annot() function. Parameter options are: x, y, w, h, fill, and ref_lbls. See control_coxph_annot() for details.

legend_pos
(numeric(2) or NULL)
vector containing x- and y-coordinates, respectively, for the legend position relative to the KM plot area. If NULL (default), the legend is positioned in the bottom right corner of the plot, or the middle right of the plot if needed to prevent overlapping.

rel_height_plot
(proportion)
proportion of total figure height to allocate to the Kaplan-Meier plot. Relative height of patients at risk table is then 1 - rel_height_plot. If annot_at_risk = FALSE or as_list = TRUE, this parameter is ignored.

ggtheme
(theme)
a graphical theme as provided by ggplot2 to format the Kaplan-Meier plot.

as_list
(flag)
whether the two ggplot objects should be returned as a list when annot_at_risk
= TRUE. If TRUE, a named list with two elements, plot and table, will be returned. If FALSE (default) the patients at risk table is printed below the plot via `cowplot::plot_grid()`.

draw [Deprecated] This function no longer generates grob objects.
newpage [Deprecated] This function no longer generates grob objects.
gp [Deprecated] This function no longer generates grob objects.
vp [Deprecated] This function no longer generates grob objects.
name [Deprecated] This function no longer generates grob objects.
annot_coxph_ref_lbls [Deprecated] Please use the ref_lbls element of control_annot_coxph instead.
position_coxph [Deprecated] Please use the x and y elements of control_annot_coxph instead.
position_surv_med [Deprecated] Please use the x and y elements of control_annot_surv_med instead.
width_annots [Deprecated] Please use the w element of control_annot_surv_med (for surv_med) and control_annot_coxph (for coxph)."

Value
A ggplot Kaplan-Meier plot and (optionally) summary table.

Examples

```r
library(dplyr)
library(nestcolor)

df <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(is_event = CNSR == 0)
variables <- list(tte = "AVAL", is_event = "is_event", arm = "ARMCD")

# Basic examples
g_km(df = df, variables = variables)
g_km(df = df, variables = variables, yval = "Failure")

# Examples with customization parameters applied
g_km(
  df = df,
  variables = variables,
  control_surv = control_surv_timepoint(conf_level = 0.9),
  col = c("grey25", "grey50", "grey75"),
  annot_at_risk_title = FALSE,
  lty = 1:3,
  font_size = 8
)

g_km(
  df = df,
  variables = variables,
  control_surv = control_surv_timepoint(conf_level = 0.9),
  col = c("grey25", "grey50", "grey75"),
  annot_at_risk_title = FALSE,
  lty = 1:3,
  font_size = 8
)
```
variables = variables,
an註_stats = c("min", "median"),
an註_stats_vlines = TRUE,
max_time = 3000,
ggtheme = ggplot2::theme_minimal()
)

# Example with pairwise Cox-PH analysis annotation table, adjusted annotation tables
g_km(
  df = df, variables = variables,
an註_coxph = TRUE,
control_coxph = control_coxph(pval_method = "wald", ties = "exact", conf_level = 0.99),
control_an註_coxph = control_coxph_an註(x = 0.26, w = 0.35),
control_an註_surv_med = control_surv_med_an註(x = 0.8, y = 0.9, w = 0.35)
)

---

**g_lineplot**

*Line plot with optional table*

**Description**

[Stable]

Line plot with the optional table.

**Usage**

```r
g_lineplot(
  df,
  alt_counts_df = NULL,
  variables = control_lineplot_vars(),
  mid = "mean",
  interval = "mean_ci",
  whiskers = c("mean_ci_lwr", "mean_ci_upr"),
  table = NULL,
  sfun = s_summary,
  ...
  mid_type = "pl",
  mid_point_size = 2,
  position = ggplot2::position_dodge(width = 0.4),
  legend_title = NULL,
  legend_position = "bottom",
  ggtheme = nestcolor::theme_nest(),
  xticks = NULL,
  xlim = NULL,
  ylim = NULL,
  x_lab = obj_label(df[[variables["x"]]]),
  y_lab = NULL,
)```
y_lab_add_paramcd = TRUE,
y_lab_add_unit = TRUE,
title = "Plot of Mean and 95% Confidence Limits by Visit",
subtitle = "",
subtitle_add_paramcd = TRUE,
subtitle_add_unit = TRUE,
caption = NULL,
table_format = get_formats_from_stats(table),
table_labels = get_labels_from_stats(table),
table_font_size = 3,
newpage = lifecycle::deprecated(),
col = NULL
}

Arguments

df (data.frame)
data set containing all analysis variables.

alt_counts_df (data.frame or NULL)
data set that will be used (only) to counts objects in groups for stratification.

variables (named character) vector of variable names in df which should include:

- x (string)
  name of x-axis variable.
- y (string)
  name of y-axis variable.
- group_var (string or NULL)
  name of grouping variable (or strata), i.e. treatment arm. Can be NA to indicate lack of groups.
- subject_var (string or NULL)
  name of subject variable. Only applies if group_var is not NULL.
- paramcd (string or NA)
  name of the variable for parameter's code. Used for y-axis label and plot's subtitle. Can be NA if paramcd is not to be added to the y-axis label or subtitle.
- y_unit (string or NA)
  name of variable with units of y. Used for y-axis label and plot's subtitle. Can be NA if y unit is not to be added to the y-axis label or subtitle.
- facet_var (string or NA)
  name of the secondary grouping variable used for plot faceting, i.e. treatment arm. Can be NA to indicate lack of groups.

mid (character or NULL)
names of the statistics that will be plotted as midpoints. All the statistics indicated in mid variable must be present in the object returned by sfun, and be of a double or numeric type vector of length one.

interval (character or NULL)
names of the statistics that will be plotted as intervals. All the statistics indicated
in interval variable must be present in the object returned by sfun, and be of a double or numeric type vector of length two. Set $\text{interval} = \text{NULL}$ if intervals should not be added to the plot.

whiskers (character)
names of the interval whiskers that will be plotted. Names must match names of the list element interval that will be returned by sfun (e.g. mean_ci_lwr element of sfun(x)[["mean_ci"]]). It is possible to specify one whisker only, or to suppress all whiskers by setting $\text{interval} = \text{NULL}$.

table (character or NULL)
names of the statistics that will be displayed in the table below the plot. All the statistics indicated in $\text{table}$ variable must be present in the object returned by sfun.

sfun (function)
the function to compute the values of required statistics. It must return a named list with atomic vectors. The names of the list elements refer to the names of the statistics and are used by mid, interval, table. It must be able to accept as input a vector with data for which statistics are computed.

... optional arguments to sfun.

mid_type (string)
controls the type of the mid plot, it can be point ("p"), line ("l"), or point and line ("pl").

mid_point_size (numeric(1))
font size of the mid plot points.

position (character or call)
geom element position adjustment, either as a string, or the result of a call to a position adjustment function.

legend_title (string)
legend title.

legend_position (string)
the position of the plot legend ("none", "left", "right", "bottom", "top", or a two-element numeric vector).

ggtheme (theme)
a graphical theme as provided by ggplot2 to control styling of the plot.

xticks (numeric or NULL)
numeric vector of tick positions or a single number with spacing between ticks on the x-axis, for use when variables$x$ is numeric. If NULL (default), \texttt{labeling::extended()} is used to determine optimal tick positions on the x-axis. If variables$x$ is not numeric, this argument is ignored.

xlim (numeric(2))
vector containing lower and upper limits for the x-axis, respectively. If NULL (default), the default scale range is used.

ylim (numeric(2))
vector containing lower and upper limits for the y-axis, respectively. If NULL (default), the default scale range is used.
x_lab (string or NULL)
x-axis label. If NULL then no label will be added.

y_lab (string or NULL)
y-axis label. If NULL then no label will be added.

y_lab_add_paramcd
(flag)
whether paramcd, i.e. unique(df[[variables["paramcd"]]]) should be added to the y-axis label (y_lab).

y_lab_add_unit (flag)
whether y-axis unit, i.e. unique(df[[variables["y_unit"]]]) should be added to the y-axis label (y_lab).

title (string)
plot title.

subtitle (string)
plot subtitle.

subtitle_add_paramcd
(flag)
whether paramcd, i.e. unique(df[[variables["paramcd"]]]) should be added to the plot's subtitle (subtitle).

subtitle_add_unit (flag)
whether the y-axis unit, i.e. unique(df[[variables["y_unit"]]]) should be added to the plot's subtitle (subtitle).

caption (string)
optional caption below the plot.

table_format (named character or NULL)
format patterns for descriptive statistics used in the (optional) table appended to the plot. It is passed directly to the h_format_row function through the format parameter. Names of table_format must match the names of statistics returned by sfun function.

table_labels (named character or NULL)
labels for descriptive statistics used in the (optional) table appended to the plot. Names of table_labels must match the names of statistics returned by sfun function.

table_font_size (numeric(1))
font size of the text in the table.

newpage [Deprecated] not used.

col (character)
color(s).

Value
A ggplot line plot (and statistics table if applicable).
Examples

```r
library(nestcolor)

# Loading data
adsl <- tern_ex_adsl
adlb <- tern_ex_adlb %>% dplyr::filter(ANL01FL == "Y", PARAMCD == "ALT", AVISIT != "SCREENING")
adlb$AVISIT <- droplevels(adlb$AVISIT)
adlb <- dplyr::mutate(adlb, AVISIT = forcats::fct_reorder(AVISIT, AVISITN, min))

# Mean with CI
g_lineplot(adlb, adsl, subtitle = "Laboratory Test:")

# Mean with CI, no stratification with group_var
g_lineplot(adlb, variables = control_lineplot_vars(group_var = NA))

# Mean, upper whisker of CI, no group_var(strata) counts N
g_lineplot(
  adlb,
  whiskers = "mean_ci_upr",
  title = "Plot of Mean and Upper 95% Confidence Limit by Visit"
)

# Median with CI
mid = "median",
interval = "median_ci",
whiskers = c("median_ci_lwr", "median_ci_upr"),
title = "Plot of Median and 95% Confidence Limits by Visit"

g_lineplot(
  adlb,
  adsl,
  mid = "median",
  interval = "median_ci",
  whiskers = c("median_ci_lwr", "median_ci_upr"),
  title = "Plot of Median and 95% Confidence Limits by Visit"
)

# Mean, +/- SD
interval = "mean_sdi",
whiskers = c("mean_sdi_lwr", "mean_sdi_upr"),
title = "Plot of Median +/- SD by Visit"

g_lineplot(adlb, adsl,
interval = "mean_sdi",
whiskers = c("mean_sdi_lwr", "mean_sdi_upr"),
title = "Plot of Median +/- SD by Visit"
)

# Mean with CI plot with stats table
adlb_f <- dplyr::filter(adlb, ARMCD != "ARM A" | AVISIT == "BASELINE")

# Mean with CI, table and customized confidence level
control = control_analyze_vars(conf_level = 0.80),
title = "Plot of Mean and 80% Confidence Limits by Visit"

g_lineplot(adlb, adsl,
  table = c("n", "mean", "mean_ci"),
  control = control_analyze_vars(conf_level = 0.80),
  title = "Plot of Mean and 80% Confidence Limits by Visit"
)
```

g_lineplot(adlb_f, table = c("n", "mean"))

---

### Description

**[Stable]**

Based on the STEP results, creates a ggplot graph showing the estimated HR or OR along the continuous biomarker value subgroups.

### Usage

```r
g_step(df, use_percentile = "Percentile Center" %in% names(df), est = list(col = "blue", lty = 1), ci_ribbon = list(fill = getOption("ggplot2.discrete.colour")[1], alpha = 0.5), col = getOption("ggplot2.discrete.colour")
```

### Arguments

- **df** (tibble): result of `tidy.step()`.
- **use_percentile** (flag): whether to use percentiles for the x axis or actual biomarker values.
- **est** (named list): `col` and `lty` settings for estimate line.
- **ci_ribbon** (named list or NULL): `fill` and `alpha` settings for the confidence interval ribbon area, or `NULL` to not plot a CI ribbon.
- **col** (character): color(s).

### Value

A ggplot STEP graph.

### See Also

Custom tidy method `tidy.step()`.
Examples

```r
g_step

library(nestcolor)
library(survival)
lung$sex <- factor(lung$sex)

# Survival example.
vars <- list(
  time = "time",
  event = "status",
  arm = "sex",
  biomarker = "age"
)

step_matrix <- fit_survival_step(
  variables = vars,
  data = lung,
  control = c(control_coxph(), control_step(num_points = 10, degree = 2))
)
step_data <- broom::tidy(step_matrix)

# Default plot.
g_step(step_data)

# Add the reference 1 horizontal line.
library(ggplot2)
g_step(step_data) +
  ggplot2::geom_hline(ggplot2::aes(yintercept = 1), linetype = 2)

# Use actual values instead of percentiles, different color for estimate and no CI,
# use log scale for y axis.
g_step(
  step_data,
  use_percentile = FALSE,
  est = list(col = "blue", lty = 1),
  ci_ribbon = NULL
) + scale_y_log10()

# Adding another curve based on additional column.
step_data$extra <- exp(step_data$"Percentile Center")
g_step(step_data) +
  ggplot2::geom_line(ggplot2::aes(y = extra), linetype = 2, color = "green")

# Response example.
vars <- list(
  response = "status",
  arm = "sex",
  biomarker = "age"
)

step_matrix <- fit_rsp_step(
  variables = vars,
  data = lung,
)
control = c(
    control_logistic(response_definition = "I(response == 2)"),
    control_step()
)
step_data <- broom::tidy(step_matrix)
g_step(step_data)

---

**g_waterfall**  
**Horizontal waterfall plot**

**Description**

[Stable]
This basic waterfall plot visualizes a quantity height ordered by value with some markup.

**Usage**

```r
l

```

**Arguments**

- **height** (numeric)  
  vector containing values to be plotted as the waterfall bars.
- **id** (character)  
  vector containing identifiers to use as the x-axis label for the waterfall bars.
- **col_var** (factor, character, or NULL)  
  categorical variable for bar coloring. NULL by default.
- **col** (character)  
  color(s).
- **xlab** (string)  
  x label. Default is "ID".
- **ylab** (string)  
  y label. Default is "Value".
**col_legend_title**

(string)

text to be displayed as legend title.

**title**

(string)

text to be displayed as plot title.

**Value**

A ggplot waterfall plot.

**Examples**

```r
library(dplyr)
library(nestcolor)

g_waterfall(height = c(3, 5, -1), id = letters[1:3])

g_waterfall(
  height = c(3, 5, -1),
  id = letters[1:3],
  col_var = letters[1:3]
)

adsl_f <- tern_ex_adsl %>%
  select(USUBJID, STUDYID, ARM, ARMCD, SEX)

adrs_f <- tern_ex_adrs %>%
  filter(PARAMCD == "OVRINV") %>%
  mutate(pchg = rnorm(n(), 10, 50))

adrs_f <- head(adrs_f, 30)
adrs_f <- adrs_f[!duplicated(adrs_f$USUBJID), ]
head(adrs_f)

g_waterfall(
  height = adrs_f$pchg,
  id = adrs_f$USUBJID,
  col_var = adrs_f$AVAlC
)

g_waterfall(
  height = adrs_f$pchg,
  id = paste("asdfsdfsdfsdf", adrs_f$USUBJID),
  col_var = adrs_f$SEX
)

g_waterfall(
  height = adrs_f$pchg,
  id = paste("asdfsdfsdfsdf", adrs_f$USUBJID),
  xlab = "ID",
  ylab = "Percentage Change",
  title = "Waterfall plot"
)
Helper function to prepare ADLB for count_abnormal_by_worst_grade().

The following pre-processing steps are applied:

1. adlb is filtered on variable avisit to only include post-baseline visits.
2. adlb is filtered on variables worst_flag_low and worst_flag_high so that only worst grades (in either direction) are included.
3. From the standard lab grade variable atoxgr, the following two variables are derived and added to adlb:
   - A grade direction variable (e.g. GRADE_DIR). The variable takes value "HIGH" when atoxgr > 0, "LOW" when atoxgr < 0, and "ZERO" otherwise.
   - A toxicity grade variable (e.g. GRADE_ANL) where all negative values from atoxgr are replaced by their absolute values.
4. Unused factor levels are dropped from adlb via droplevels().

Usage

h_adlb_abnormal_by_worst_grade(
  adlb,
  atoxgr = "ATOXGR",
  avisit = "AVISIT",
  worst_flag_low = "WGRLOFL",
  worst_flag_high = "WGRHIFL"
)

Arguments

- adlb: (data.frame) ADLB data frame.
- atoxgr: (string) name of the analysis toxicity grade variable. This must be a factor variable.
- avisit: (string) name of the analysis visit variable.
**h_adlb_worsen**

- **worst_flag_low** (string)
  - name of the worst low lab grade flag variable. This variable is set to "Y" when indicating records of worst low lab grades.

- **worst_flag_high** (string)
  - name of the worst high lab grade flag variable. This variable is set to "Y" when indicating records of worst high lab grades.

**Value**

Function `h_adlb_abnormal_by_worst_grade()` returns the adlb data frame with two new variables: `GRADE_DIR` and `GRADE_ANL`.

**See Also**

- `abnormal_by_worst_grade`

**Examples**

```r
h_adlb_abnormal_by_worst_grade(tern_ex_adlb) %>%
dplyr::select(ATOXGR, GRADE_DIR, GRADE_ANL) %>%
head(10)
```

---

**Description**

**[Stable]**

Helper function to prepare a df for generate the patient count shift table.

**Usage**

```r
h_adlb_worsen(
  adlb,
  worst_flag_low = NULL,
  worst_flag_high = NULL,
  direction_var
)
```

**Arguments**

- **adlb** (data.frame)
  - ADLB data frame.

- **worst_flag_low** (named vector)
  - worst low post-baseline lab grade flag variable. See how this is implemented in the following examples.
worst_flag_high
(named vector)
worst high post-baseline lab grade flag variable. See how this is implemented in
the following examples.
direction_var (string)
name of the direction variable specifying the direction of the shift table of inter-
est. Only lab records flagged by L, H or B are included in the shift table.
  • L: low direction only
  • H: high direction only
  • B: both low and high directions

Value

h_adlb_worsen() returns the adlb data.frame containing only the worst labs specified according
to worst_flag_low or worst_flag_high for the direction specified according to direction_var.
For instance, for a lab that is needed for the low direction only, only records flagged by worst_flag_low
are selected. For a lab that is needed for both low and high directions, the worst low records are
selected for the low direction, and the worst high record are selected for the high direction.

See Also

abnormal_by_worst_grade_worsen

Examples

library(dplyr)

# The direction variable, GRADDR, is based on metadata
adlb < tern_ex_adlb %>%
  mutate(
    GRADDR = case_when(
      PARAMCD == "ALT" ~ "B",
      PARAMCD == "CRP" ~ "L",
      PARAMCD == "IGA" ~ "H"
    )
  ) %>%
  filter(SAFFL == "Y" & ONTRTFL == "Y" & GRADDR != "")

df <- h_adlb_worsen(
adlb,
worst_flag_low = c("WGRLOFL" = "Y"),
worst_flag_high = c("WGRHIFL" = "Y"),
direction_var = "GRADDR"
)
h_adsl_adlb_merge_using_worst_flag

Helper function for deriving analysis datasets for select laboratory tables

Description

[Stable]

Helper function that merges ADSL and ADLB datasets so that missing lab test records are inserted in the output dataset. Remember that na_level must match the needed pre-processing done with df_explicit_na() to have the desired output.

Usage

```
h_adsl_adlb_merge_using_worst_flag(
  adsl,
  adlb,
  worst_flag = c(WGRHIFL = "Y"),
  by_visit = FALSE,
  no_fillin_visits = c("SCREENING", "BASELINE")
)
```

Arguments

- **adsl** (data.frame)
  - ADSL data frame.
- **adlb** (data.frame)
  - ADLB data frame.
- **worst_flag** (named character)
  - worst post-baseline lab flag variable. See how this is implemented in the following examples.
- **by_visit** (flag)
  - defaults to FALSE to generate worst grade per patient. If worst grade per patient per visit is specified for worst_flag, then by_visit should be TRUE to generate worst grade patient per visit.
- **no_fillin_visits** (named character)
  - visits that are not considered for post-baseline worst toxicity grade. Defaults to c("SCREENING", "BASELINE").

Details

In the result data missing records will be created for the following situations:

- Patients who are present in adsl but have no lab data in adlb (both baseline and post-baseline).
- Patients who do not have any post-baseline lab values.
- Patients without any post-baseline values flagged as the worst.
Value
df containing variables shared between adlb and adsl along with variables PARAM, PARAMCD, ATOXGR, and BTOXGR relevant for analysis. Optionally, AVISIT are AVISITN are included when by_visit = TRUE and no_fillin_visits = c("SCREENING", "BASELINE").

Examples

```r
# `h_adsl_adlb_merge_using_worst_flag`
adlb_out <- h_adsl_adlb_merge_using_worst_flag(
  tern_ex_adsl,
  tern_ex_adlb,
  worst_flag = c("WGRHIFL" = "Y")
)

# `h_adsl_adlb_merge_using_worst_flag` by visit example
adlb_out_by_visit <- h_adsl_adlb_merge_using_worst_flag(
  tern_ex_adsl,
  tern_ex_adlb,
  worst_flag = c("WGRLOVFL" = "Y"),
  by_visit = TRUE
)
```

---

**h_ancova**

Helper function to return results of a linear model

Description

[Stable]

Usage

`h_ancova(.var, .df_row, variables, interaction_item = NULL)`

Arguments

- `.var` *(string)*
  single variable name that is passed by rtables when requested by a statistics function.
- `.df_row` *(data.frame)*
  data set that includes all the variables that are called in `.var` and `variables`.
- `variables` *(named list of string)*
  list of additional analysis variables, with expected elements:
  - `arm` *(string)*
    group variable, for which the covariate adjusted means of multiple groups will be summarized. Specifically, the first level of `arm` variable is taken as the reference group.
covariates (character)

a vector that can contain single variable names (such as "X1"), and/or interaction terms indicated by "X1 * X2".

interaction_item

(string or NULL)

name of the variable that should have interactions with arm. if the interaction is not needed, the default option is NULL.

Value

The summary of a linear model.

Examples

h_ancova(
  .var = "Sepal.Length",
  .df_row = iris,
  variables = list(arm = "Species", covariates = c("Petal.Length * Petal.Width", "Sepal.Width"))
)

h_append_grade_groups  Helper function for s_count_occurrences_by_grade()
remove_single (flag)
  TRUE to not include the elements of one-element grade groups in the output
  list; in this case only the grade groups names will be included in the output. If
  only_grade_groups is set to TRUE this argument is ignored.

only_grade_groups (flag)
  whether only the specified grade groups should be included, with individual
  grade rows removed (TRUE), or all grades and grade groups should be displayed
  (FALSE).

Value
  Formatted list of grade groupings.

Examples

h_append_grade_groups(
  list(
    "Any Grade" = as.character(1:5),
    "Grade 1-2" = c("1", "2"),
    "Grade 3-4" = c("3", "4")
  ),
  list("1" = 10, "2" = 20, "3" = 30, "4" = 40, "5" = 50)
)

h_append_grade_groups(
  list(
    "Any Grade" = as.character(5:1),
    "Grade A" = "5",
    "Grade B" = c("4", "3")
  ),
  list("1" = 10, "2" = 20, "3" = 30, "4" = 40, "5" = 50)
)

h_append_grade_groups(
  list(
    "Any Grade" = as.character(1:5),
    "Grade 1-2" = c("1", "2"),
    "Grade 3-4" = c("3", "4")
  ),
  list("1" = 10, "2" = 5, "3" = 0)
)
**h_count_cumulative**

**Description**

[Stable]
Helper function to extract column indices from a VTableTree for a given vector of column names.

**Usage**

```r
h_col_indices(table_tree, col_names)
```

**Arguments**

- `table_tree` (VTableTree)
  - rtables table object to extract the indices from.

- `col_names` (character)
  - vector of column names.

**Value**

A vector of column indices.

---

**h_count_cumulative**  
*Helper function for s_count_cumulative()*

**Description**

[Stable]
Helper function to calculate count and fraction of x values in the lower or upper tail given a threshold.

**Usage**

```r
h_count_cumulative(  
  x,  
  threshold,  
  lower_tail = TRUE,  
  include_eq = TRUE,  
  na.rm = TRUE,  
  .N_col
)
```

**Arguments**

- `x` (numeric)
  - vector of numbers we want to analyze.

- `threshold` (numeric(1))
  - a cutoff value as threshold to count values of x.
lower_tail (flag)
whether to count lower tail, default is TRUE.

include_eq (flag)
whether to include value equal to the threshold in count, default is TRUE.

na.rm (flag)
whether NA values should be removed from x prior to analysis.

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typically passed by rtables.

Value
A named vector with items:

- count: the count of values less than, less or equal to, greater than, or greater or equal to a threshold of user specification.
- fraction: the fraction of the count.

See Also

count_cumulative

Examples

set.seed(1, kind = "Mersenne-Twister")
x <- c(sample(1:10, 10), NA)
.N_col <- length(x)

h_count_cumulative(x, 5, .N_col = .N_col)
h_count_cumulative(x, 5, lower_tail = FALSE, include_eq = FALSE, na.rm = FALSE, .N_col = .N_col)
h_count_cumulative(x, 0, lower_tail = FALSE, .N_col = .N_col)
h_count_cumulative(x, 100, lower_tail = FALSE, .N_col = .N_col)
Usage

h_coxreg_univar_formulas(variables, interaction = FALSE)

h_coxreg_multivar_formula(variables)

h_coxreg_univar_extract(effect, covar, data, mod, control = control_coxreg())

h_coxreg_multivar_extract(var, data, mod, control = control_coxreg())

Arguments

variables (named list of string)
list of additional analysis variables.

interaction (flag)
if TRUE, the model includes the interaction between the studied treatment and candidate covariate. Note that for univariate models without treatment arm, and multivariate models, no interaction can be used so that this needs to be FALSE.

effect (string)
the treatment variable.

covar (string)
the name of the covariate in the model.

data (data.frame)
the dataset containing the variables to summarize.

mod (coxph)
Cox regression model fitted by survival::coxph().

control (list)
a list of controls as returned by control_coxreg().

var (string)
single variable name that is passed by rtables when requested by a statistics function.

Value

- h_coxreg_univar_formulas() returns a character vector coercible into formulas (e.g stats::as.formula()).

- h_coxreg_multivar_formula() returns a string coercible into a formula (e.g stats::as.formula()).

- h_coxreg_univar_extract() returns a data.frame with variables effect, term, term_label, level, n, hr, lcl, ucl, and pval.

- h_coxreg_multivar_extract() returns a data.frame with variables pval, hr, lcl, ucl, level, n, term, and term_label.
Functions

- **h_coxreg_univar_formulas()**: Helper for Cox regression formula. Creates a list of formulas. It is used internally by `fit_coxreg_univar()` for the comparison of univariate Cox regression models.

- **h_coxreg_multivar_formula()**: Helper for multivariate Cox regression formula. Creates a formulas string. It is used internally by `fit_coxreg_multivar()` for the comparison of multivariate Cox regression models. Interactions will not be included in multivariate Cox regression model.

- **h_coxreg_univar_extract()**: Utility function to help tabulate the result of a univariate Cox regression model.

- **h_coxreg_multivar_extract()**: Tabulation of multivariate Cox regressions. Utility function to help tabulate the result of a multivariate Cox regression model for a treatment/covariate variable.

See Also

cox_regression

Examples

```r
# 'h_coxreg_univar_formulas'

## Simple formulas.
h_coxreg_univar_formulas(
    variables = list(
        time = "time", event = "status", arm = "armcd", covariates = c("X", "y")
    )
)

## Addition of an optional strata.
h_coxreg_univar_formulas(
    variables = list(
        time = "time", event = "status", arm = "armcd", covariates = c("X", "y"),
        strata = "SITE"
    )
)

## Inclusion of the interaction term.
h_coxreg_univar_formulas(
    variables = list(
        time = "time", event = "status", arm = "armcd", covariates = c("X", "y"),
        strata = "SITE",
        interaction = TRUE
    )
)

## Only covariates fitted in separate models.
h_coxreg_univar_formulas(
    variables = list(
        time = "time", event = "status", covariates = c("X", "y")
    )
)
```
h_coxxreg_multivar_formula

```r
h_coxxreg_multivar_formula(
  variables = list(
    time = "AVAL", event = "event", arm = "ARMCD", covariates = c("RACE", "AGE")
  )
)
```

# Addition of an optional strata.
```r
h_coxxreg_multivar_formula(
  variables = list(
    time = "AVAL", event = "event", arm = "ARMCD", covariates = c("RACE", "AGE"),
    strata = "SITE"
  )
)
```

# Example without treatment arm.
```r
h_coxxreg_multivar_formula(
  variables = list(
    time = "AVAL", event = "event", covariates = c("RACE", "AGE"),
    strata = "SITE"
  )
)
```

library(survival)

dta_simple <- data.frame(
  time = c(5, 5, 10, 10, 5, 5, 10, 10),
  status = c(0, 0, 1, 0, 0, 1, 1, 1),
  armcd = factor(LETTERS[c(1, 1, 1, 1, 2, 2, 2, 2)], levels = c("A", "B")),
  var1 = c(45, 55, 65, 75, 55, 65, 85, 75),
  var2 = c("F", "M", "F", "M", "F", "M", "F", "U")
)

mod <- coxph(Surv(time, status) ~ armcd + var1, data = dta_simple)
result <- h_coxxreg_univar_extract(
  effect = "armcd", covar = "armcd", mod = mod, data = dta_simple
)
result

mod <- coxph(Surv(time, status) ~ armcd + var1, data = dta_simple)
result <- h_coxxreg_multivar_extract(
  var = "var1", mod = mod, data = dta_simple
)
result
**Description**

[Stable]

Convert the survival fit data into a data frame designed for plotting within `g_km`.
This starts from the `broom::tidy()` result, and then:

- Post-processes the `strata` column into a factor.
- Extends each stratum by an additional first row with time 0 and probability 1 so that downstream plot lines start at those coordinates.
- Adds a `censor` column.
- Filters the rows before `max_time`.

**Usage**

```r
h_data_plot(fit_km, armval = "All", max_time = NULL)
```

**Arguments**

- `fit_km` (survfit): result of `survival::survfit()`.
- `armval` (string): used as strata name when treatment arm variable only has one level. Default is "All".
- `max_time` (numeric(1)): maximum value to show on x-axis. Only data values less than or up to this threshold value will be plotted (defaults to `NULL`).

**Value**

A tibble with columns `time`, `n.risk`, `n.event`, `n.censor`, `estimate`, `std.error`, `conf.high`, `conf.low`, `strata`, and `censor`.

**Examples**

```r
library(dplyr)
library(survival)

# Test with multiple arms
tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .) %>%
  h_data_plot()

# Test with single arm
tern_ex_adtte %>%
  filter(PARAMCD == "OS", ARMCD == "ARM B") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .) %>%
  h_data_plot(armval = "ARM B")
```
Description

[Deprecated]
The elements composing the ggplot are extracted and organized in a list.

Usage

\texttt{h\_decompose\_gg(gg)}

Arguments

\texttt{gg} \hspace{1cm} (ggplot)
a graphic to decompose.

Value

A named list with elements:

- \texttt{panel}: The panel.
- \texttt{yaxis}: The y-axis.
- \texttt{xaxis}: The x-axis.
- \texttt{xlab}: The x-axis label.
- \texttt{ylab}: The y-axis label.
- \texttt{guide}: The legend.

Examples

\begin{verbatim}
library(dplyr)
library(survival)
library(grid)
fit_km <- tern_ex_adtte %>%
    filter(PARAMCD == "OS") %>%
    survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .)
data_plot <- h\_data\_plot(fit_km = fit_km)
xticks <- h\_xticks(data = data_plot)
gg <- h\_ggkm(
    data = data_plot,
yval = "Survival",
censor\_show = TRUE,
xticks = xticks, xlab = "Days", ylab = "Survival Probability",
title = "tt",
footnotes = "ff"
)
\end{verbatim}
```r
g_el <- h_decompose.gg(gg)
grid::grid.newpage()
grid.rect(gp = grid::gpar(lty = 1, col = "red", fill = "gray85", lwd = 5))
grid::grid.draw(g_el$panel)

grid::grid.newpage()
grid.rect(gp = grid::gpar(lty = 1, col = "royalblue", fill = "gray85", lwd = 5))
grid::grid.draw(with(g_el, cbind(ylab, yaxis)))
```

---

### h_format_row

Helper function to format the optional g_lineplot table

---

**Description**

[Stable]

**Usage**

```r
h_format_row(x, format, labels = NULL)
```

**Arguments**

- **x**  
  (named list)  
  list of numerical values to be formatted and optionally labeled. Elements of x must be numeric vectors.

- **format**  
  (named character or NULL)  
  format patterns for x. Names of the format must match the names of x. This parameter is passed directly to the rtables::format_rcell function through the format parameter.

- **labels**  
  (named character or NULL)  
  optional labels for x. Names of the labels must match the names of x. When a label is not specified for an element of x, then this function tries to use label or names (in this order) attribute of that element (depending on which one exists and it is not NULL or NA or NaN). If none of these attributes are attached to a given element of x, then the label is automatically generated.

**Value**

A single row data.frame object.
Examples

```r
mean_ci <- c(48, 51)
x <- list(mean = 50, mean_ci = mean_ci)
format <- c(mean = "xx.x", mean_ci = "(xx.xx, xx.xx)")
labels <- c(mean = "My Mean")
h_format_row(x, format, labels)

attr(mean_ci, "label") <- "Mean 95% CI"
x <- list(mean = 50, mean_ci = mean_ci)
h_format_row(x, format, labels)
```

Description

[Deprecated]
Draw the Kaplan-Meier plot using ggplot2.

Usage

```r
h_ggkm(
data,
xticks = NULL,
yval = "Survival",
censor_show,
xlab,
ylab,
ylim = NULL,
title,
footnotes = NULL,
max_time = NULL,
lwd = 1,
lty = NULL,
pch = 3,
size = 2,
col = NULL,
ci_ribbon = FALSE,
ggtheme = nestcolor::theme_nest()
)
```

Arguments

data (data.frame)
survival data as pre-processed by h_data_plot.
xticks (numeric or NULL)
numeric vector of tick positions or a single number with spacing between ticks on the x-axis. If NULL (default), `labeling::extended()` is used to determine optimal tick positions on the x-axis.

yval (string)
type of plot, to be plotted on the y-axis. Options are Survival (default) and Failure probability.

censor_show (flag)
whether to show censored observations.

xlab (string)
x-axis label.

ylab (string)
y-axis label.

ylim (numeric(2))
vector containing lower and upper limits for the y-axis, respectively. If NULL (default), the default scale range is used.

title (string)
plot title.

footnotes (string)
plot footnotes.

max_time (numeric(1))
maximum value to show on x-axis. Only data values less than or up to this threshold value will be plotted (defaults to NULL).

lwd (numeric)
line width. If a vector is given, its length should be equal to the number of strata from `survival::survfit()`.

lty (numeric)
line type. If a vector is given, its length should be equal to the number of strata from `survival::survfit()`.

pch (string)
name of symbol or character to use as point symbol to indicate censored cases.

size (numeric(1))
size of censored point symbols.

col (character)
lines colors. Length of a vector should be equal to number of strata from `survival::survfit()`.

ci_ribbon (flag)
whether the confidence interval should be drawn around the Kaplan-Meier curve.

ggtheme (theme)
a graphical theme as provided by `ggplot2` to format the Kaplan-Meier plot.

Value

A ggplot object.
Examples

```r
library(dplyr)
library(survival)

fit_km <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .)
data_plot <- h_data_plot(fit_km = fit_km)
xticks <- h_xticks(data = data_plot)
gg <- h_ggkm(
data = data_plot,
censor_show = TRUE,
xticks = xticks,
xlab = "Days",
yval = "Survival",
ylab = "Survival Probability",
title = "Survival"
)
gg
```

---

**h_grob_coxph**  
*Helper function to create Cox-PH grobs*

**Description**

[Deprecated]

Grob of `rtable` output from `h_tbl_coxph_pairwise()`

**Usage**

```r
h_grob_coxph(
  ..., 
  x = 0,
  y = 0,
  width = grid::unit(0.4, "npc"),
ttheme = gridExtra::ttheme_default(padding = grid::unit(c(1, 0.5), "lines"), core = 
    list(bg_params = list(fill = c("grey95", "grey90"), alpha = 0.5)))
)
```

**Arguments**

- `...`: arguments to pass to `h_tbl_coxph_pairwise()`.
- `x` (proportion): a value between 0 and 1 specifying x-location.
- `y` (proportion): a value between 0 and 1 specifying y-location.
width (grid::unit)
width (as a unit) to use when printing the grob.

ttheme (list)
see gridExtra::ttheme_default().

Value
A grob of a table containing statistics HR, XX% CI (XX taken from control_coxph_pw), and p-value (log-rank).

Examples

```r
library(dplyr)
library(survival)
library(grid)

grid::grid.newpage()
grid.rect(gp = grid::gpar(lty = 1, col = "pink", fill = "gray85", lwd = 1))
data <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(is_event = CNSR == 0)
tbl_grob <- h_grob_coxph(
  df = data,
  variables = list(tte = "AVAL", is_event = "is_event", arm = "ARMCD"),
  control_coxph_pw = control_coxph(conf_level = 0.9), x = 0.5, y = 0.5
)
grid::grid.draw(tbl_grob)
```

h_grob_median_surv

*Helper function to create survival estimation grobs*

Description

[Deprecated]
The survival fit is transformed in a grob containing a table with groups in rows characterized by N, median and 95% confidence interval.

Usage

```r
h_grob_median_surv(
  fit_km,
  armval = "All",
  x = 0.9,
  y = 0.9,
  width = grid::unit(0.3, "npc"),
  ttheme = gridExtra::ttheme_default()
)
```
Arguments

- **fit_km** (survfit) result of `survival::survfit()`.
- **armval** (string) used as strata name when treatment arm variable only has one level. Default is "All".
- **x** (proportion) a value between 0 and 1 specifying x-location.
- **y** (proportion) a value between 0 and 1 specifying y-location.
- **width** (grid::unit) width (as a unit) to use when printing the grob.
- **ttheme** (list) see `gridExtra::ttheme_default()`.

Value

A grob of a table containing statistics N, Median, and XX% CI (XX taken from fit_km).

Examples

```r
library(dplyr)
library(survival)
library(grid)

grid::grid.newpage()
grid.rect(gp = grid::gpar(lty = 1, col = "pink", fill = "gray85", lwd = 1))
tern_ex_adtte %%
  filter(PARAMCD == "OS") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .) %>%
  h_grob_median_surv() %>%
  grid::grid.draw()
```

---

### h_grob_tbl_at_risk

*Helper function to create patient-at-risk grobs*

### Description

[Deprecated]

Two graphical objects are obtained, one corresponding to row labeling and the second to the table of numbers of patients at risk. If `title` = TRUE, a third object corresponding to the table title is also obtained.
Usage

h_grob_tbl_at_risk(data, annot_tbl, xlim, title = TRUE)

Arguments

data (data.frame)
survival data as pre-processed by h_data_plot.

annot_tbl (data.frame)
annotation as prepared by survival::summary.survfit() which includes the number of patients at risk at given time points.

xlim (numeric(1))
the maximum value on the x-axis (used to ensure the at risk table aligns with the KM graph).

title (flag)
whether the "Patients at Risk" title should be added above the annot_at_risk table. Has no effect if annot_at_risk is FALSE. Defaults to TRUE.

Value

A named list of two gTree objects if title = FALSE: at_risk and label, or three gTree objects if title = TRUE: at_risk, label, and title.

Examples

library(dplyr)
library(survival)
library(grid)

fit_km <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .)

data_plot <- h_data_plot(fit_km = fit_km)

xticks <- h_xticks(data = data_plot)

gg <- h_ggkm(
data = data_plot,
censor_show = TRUE,
xticks = xticks, xlab = "Days", ylab = "Survival Probability",
title = "tt", footnotes = "ff", yval = "Survival"
)

# The annotation table reports the patient at risk for a given strata and # times ('xticks').
annot_tbl <- summary(fit_km, times = xticks)
if (is.null(fit_km$strata)) {
  annot_tbl <- with(annot_tbl, data.frame(n.risk = n.risk, time = time, strata = "All"))
} else {
  strata_lst <- strsplit(sub("=" , "equals", levels(annot_tbl$strata)), "equals")
}
levels(annot_tbl$strata) <- matrix(unlist(strata_lst), ncol = 2, byrow = TRUE)[, 2]
annot_tbl <- data.frame(
  n.risk = annot_tbl$n.risk,
  time = annot_tbl$time,
  strata = annot_tbl$strata
)

# The annotation table is transformed into a grob.
tbl <- h_grob_tbl_at_risk(data = data_plot, annot_tbl = annot_tbl, xlim = max(xticks))

# For the representation, the layout is estimated for which the decomposition
# of the graphic element is necessary.
g_el <- h_decompose_gg(gg)
lyt <- h_km_layout(data = data_plot, g_el = g_el, title = "t", footnotes = "f")

grid::grid.newpage()
pushViewport(viewport(layout = lyt, height = .95, width = .95))
grid.rect(gp = grid::gpar(lty = 1, col = "purple", fill = "gray85", lwd = 1))
pushViewport(viewport(layout.pos.row = 3:4, layout.pos.col = 2))
grid.rect(gp = grid::gpar(lty = 1, col = "orange", fill = "gray85", lwd = 1))
grid::grid.draw(tbl$at_risk)
popViewport()
pushViewport(viewport(layout.pos.row = 3:4, layout.pos.col = 1))
grid.rect(gp = grid::gpar(lty = 1, col = "green3", fill = "gray85", lwd = 1))
grid::grid.draw(tbl$label)

---

**h_grob_y_annot**

Helper function to create grid object with y-axis annotation

**Description**

[Deprecated]

Build the y-axis annotation from a decomposed ggplot.

**Usage**

h_grob_y_annot(ylab, yaxis)

**Arguments**

- **ylab** (gtable)
  
  the y-lab as a graphical object derived from a ggplot.

- **yaxis** (gtable)
  
  the y-axis as a graphical object derived from a ggplot.
A gTree object containing the y-axis annotation from a ggplot.

Examples

```r
llibrary(dplyr)
llibrary(survival)
llibrary(grid)

fit_km <- tern_ex_adtte %>%
filter(PARAMCD == "OS") %>%
survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .)
data_plot <- h_data_plot(fit_km = fit_km)
xticks <- h_xticks(data = data_plot)
gg <- h_ggkm(
  data = data_plot,
censor_show = TRUE,
  xticks = xticks, xlab = "Days", ylab = "Survival Probability",
title = "title", footnotes = "footnotes", yval = "Survival"
)
g_el <- h_decompose_gg(gg)

grid::grid.newpage()
pvp <- grid::plotViewport(margins = c(5, 4, 2, 20))
pushViewport(pvp)
grid::grid.draw(h_grob_y_annot(ylab = g_el$ylab, yaxis = g_el$yaxis))
grid.rect(gp = grid::gpar(lty = 1, col = "gray35", fill = NA))
```

---

**h_g_ipp**

**Helper function to create simple line plot over time**

**Description**

[Stable]

Function that generates a simple line plot displaying parameter trends over time.

**Usage**

```r
h_g_ipp(
  df,
  xvar,
  yvar,
  xlab,
  ylab,
  id_var,
```
Arguments

df \begin{itemize} 
\item \texttt{data.frame} 
\item data set containing all analysis variables. \end{itemize}

xvar \begin{itemize} 
\item \texttt{string} 
\item time point variable to be plotted on x-axis. \end{itemize}

yvar \begin{itemize} 
\item \texttt{string} 
\item continuous analysis variable to be plotted on y-axis. \end{itemize}

xlab \begin{itemize} 
\item \texttt{string} 
\item plot label for x-axis. \end{itemize}

ylab \begin{itemize} 
\item \texttt{string} 
\item plot label for y-axis. \end{itemize}

id_var \begin{itemize} 
\item \texttt{string} 
\item variable used as patient identifier. \end{itemize}

title \begin{itemize} 
\item \texttt{string} 
\item title for plot. \end{itemize}

subtitle \begin{itemize} 
\item \texttt{string} 
\item subtitle for plot. \end{itemize}

caption \begin{itemize} 
\item \texttt{string} 
\item optional caption below the plot. \end{itemize}

add_baseline_hline \begin{itemize} 
\item \texttt{flag} 
\item adds horizontal line at baseline y-value on plot when TRUE. \end{itemize}

yvar_baseline \begin{itemize} 
\item \texttt{string} 
\item variable with baseline values only. Ignored when add_baseline_hline is FALSE. \end{itemize}

ggtheme \begin{itemize} 
\item \texttt{theme} 
\item optional graphical theme function as provided by \texttt{ggplot2} to control outlook of plot. Use \texttt{ggplot2::theme()} to tweak the display. \end{itemize}

col \begin{itemize} 
\item \texttt{character} 
\item line colors. \end{itemize}

Value

A \texttt{ggplot} line plot.

See Also

gg IPP() which uses this function.
Examples

```r
library(dplyr)
library(nestcolor)

# Select a small sample of data to plot.
adlb <- tern_ex_adlb %>%
  filter(PARAMCD == "ALT", !(AVISIT %in% c("SCREENING", "BASELINE"))) %>%
slice(1:36)

p <- h_g_ipp(
  df = adlb,
  xvar = "AVISIT",
  yvar = "AVAL",
  xlab = "Visit",
  id_var = "USUBJID",
  ylab = "SGOT/ALT (U/L)",
  add_baseline_hline = TRUE
)
p
```

---

**h_km_layout**

**Helper function to prepare a KM layout**

**Description**

[Deprecated]

Prepares a (5 rows) x (2 cols) layout for the Kaplan-Meier curve.

**Usage**

```r
h_km_layout(
  data,
  g_el,
  title,
  footnotes,
  annot_at_risk = TRUE,
  annot_at_risk_title = TRUE
)
```

**Arguments**

- `data` *(data.frame)*
  - survival data as pre-processed by `h_data_plot`.
- `g_el` *(list of gtable)*
  - list as obtained by `h_decompose_gg()`.
- `title` *(string)*
  - plot title.
footnotes (string)
plot footnotes.

annot_at_risk (flag)
compute and add the annotation table reporting the number of patient at risk matching the main grid of the Kaplan-Meier curve.

annot_at_risk_title (flag)
whether the "Patients at Risk" title should be added above the annot_at_risk table. Has no effect if annot_at_risk is FALSE. Defaults to TRUE.

Details

The layout corresponds to a grid of two columns and five rows of unequal dimensions. Most of the dimension are fixed, only the curve is flexible and will accommodate with the remaining free space.

- The left column gets the annotation of the ggplot (y-axis) and the names of the strata for the patient at risk tabulation. The main constraint is about the width of the columns which must allow the writing of the strata name.
- The right column receive the ggplot, the legend, the x-axis and the patient at risk table.

Value

A grid layout.

Examples

library(dplyr)
library(survival)
library(grid)

fit_km <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .)
data_plot <- h_data_plot(fit_km = fit_km)
xticks <- h_xticks(data = data_plot)
gg <- h_ggkm(
data = data_plot,
censor_show = TRUE,
  xticks = xticks, xlab = "Days", ylab = "Survival Probability",
title = "tt", footnotes = "ff", yval = "Survival"
}
g_el <- h_decompose_gg(gg)
lyt <- h_km_layout(data = data_plot, g_el = g_el, title = "t", footnotes = "f")
grid.show.layout(lyt)
Description

[Stable]
Helper functions used in calculations for logistic regression.

Usage

h_get_interaction_vars(fit_glm)

h_interaction_coef_name(
  interaction_vars,
  first_var_with_level,
  second_var_with_level
)

h_or_cat_interaction(
  odds_ratio_var,
  interaction_var,
  fit_glm,
  conf_level = 0.95
)

h_or_cont_interaction(
  odds_ratio_var,
  interaction_var,
  fit_glm,
  at = NULL,
  conf_level = 0.95
)

h_or_interaction(
  odds_ratio_var,
  interaction_var,
  fit_glm,
  at = NULL,
  conf_level = 0.95
)

h_simple_term_labels(terms, table)

h_interaction_term_labels(terms1, terms2, table, any = FALSE)

h_glm_simple_term_extract(x, fit_glm)
h_logistic_regression

h_glm_interaction_extract(x, fit_glm)

h_glm_inter_term_extract(odds_ratio_var, interaction_var, fit_glm, ...)

h_logistic_simple_terms(x, fit_glm, conf_level = 0.95)

h_logistic_inter_terms(x, fit_glm, conf_level = 0.95, at = NULL)

Arguments

fit_glm (glm) logistic regression model fitted by stats::glm() with "binomial" family. Limited functionality is also available for conditional logistic regression models fitted by survival::clogit(), currently this is used only by extract_rsp_biomarkers().

interaction_vars (character(2)) interaction variable names.

first_var_with_level (character(2)) the first variable name with the interaction level.

second_var_with_level (character(2)) the second variable name with the interaction level.

odds_ratio_var (string) the odds ratio variable.

interaction_var (string) the interaction variable.

conf_level (proportion) confidence level of the interval.

at (numeric or NULL) optional values for the interaction variable. Otherwise the median is used.

terms (character) simple terms.

table (table) table containing numbers for terms.

terms1 (character) terms for first dimension (rows).

terms2 (character) terms for second dimension (rows).

any (flag) whether any of term1 and term2 can be fulfilled to count the number of patients. In that case they can only be scalar (strings).

x (character) a variable or interaction term in fit_glm (depending on the helper function used).
... additional arguments for the lower level functions.

Value

Vector of names of interaction variables.
Name of coefficient.
Odds ratio.
Odds ratio.
Term labels containing numbers of patients.
Tabulated main effect results from a logistic regression model.
Tabulated interaction term results from a logistic regression model.
A data.frame of tabulated interaction term results from a logistic regression model.
Tabulated statistics for the given variable(s) from the logistic regression model.
Tabulated statistics for the given variable(s) from the logistic regression model.

Functions

- h_get_interaction_vars(): Helper function to extract interaction variable names from a fitted model assuming only one interaction term.
- h_interaction_coef_name(): Helper function to get the right coefficient name from the interaction variable names and the given levels. The main value here is that the order of first and second variable is checked in the interaction_vars input.
- h_or_cat_interaction(): Helper function to calculate the odds ratio estimates for the case when both the odds ratio and the interaction variable are categorical.
- h_or_cont_interaction(): Helper function to calculate the odds ratio estimates for the case when either the odds ratio or the interaction variable is continuous.
- h_or_interaction(): Helper function to calculate the odds ratio estimates in case of an interaction. This is a wrapper for h_or_cont_interaction() and h_or_cat_interaction().
- h_simple_term_labels(): Helper function to construct term labels from simple terms and the table of numbers of patients.
- h_interaction_term_labels(): Helper function to construct term labels from interaction terms and the table of numbers of patients.
- h_glm_simple_term_extract(): Helper function to tabulate the main effect results of a (conditional) logistic regression model.
- h_glm_interaction_extract(): Helper function to tabulate the interaction term results of a logistic regression model.
- h_glm_inter_term_extract(): Helper function to tabulate the interaction results of a logistic regression model. This basically is a wrapper for h_or_interaction() and h_glm_simple_term_extract() which puts the results in the right data frame format.
- h_logistic_simple_terms(): Helper function to tabulate the results including odds ratios and confidence intervals of simple terms.
• `h_logistic_inter_terms()`: Helper function to tabulate the results including odds ratios and confidence intervals of interaction terms.

Note

We don’t provide a function for the case when both variables are continuous because this does not arise in this table, as the treatment arm variable will always be involved and categorical.

Examples

```r
library(dplyr)
library(broom)

adrs_f <- tern_ex_adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(RACE %in% c("ASIAN", "WHITE", "BLACK OR AFRICAN AMERICAN")) %>%
  mutate(
    Response = case_when(AVALC %in% c("PR", "CR") ~ 1, TRUE ~ 0),
    RACE = factor(RACE),
    SEX = factor(SEX)
  )
formats::var_labels(adrs_f) <- c(formatters::var_labels(tern_ex_adrs), Response = "Response")
mod1 <- fit_logistic(
  data = adrs_f,
  variables = list(  
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE")
  )
)
mod2 <- fit_logistic(
  data = adrs_f,
  variables = list(  
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE"),
    interaction = "AGE"
  )
)

h_glm_simple_term_extract("AGE", mod1)

h_glm_simple_term_extract("ARMCD", mod1)

h_glm_interaction_extract("ARMCD:AGE", mod2)

h_glm_inter_term_extract("AGE", "ARMCD", mod2)

h_logistic_simple_terms("AGE", mod1)

h_logistic_inter_terms(c("RACE", "AGE", "ARMCD", "AGE:ARMCD"), mod2)
```
h_map_for_count_abnormal

Helper function to create a map data frame for trim_levels_to_map()

Description

[Stable]

Helper function to create a map data frame from the input dataset, which can be used as an argument in the trim_levels_to_map split function. Based on different method, the map is constructed differently.

Usage

h_map_for_count_abnormal(
  df,
  variables = list(anl = "ANRIND", split_rows = c("PARAM"), range_low = "ANRLO", range_high = "ANRHI"),
  abnormal = list(low = c("LOW", "LOW LOW"), high = c("HIGH", "HIGH HIGH")),
  method = c("default", "range"),
  na_str = "<Missing>"
)

Arguments

df (data.frame) data set containing all analysis variables.
variables (named list of string) list of additional analysis variables.
abnormal (named list) identifying the abnormal range level(s) in df. Based on the levels of abnormality of the input dataset, it can be something like list(Low = "LOW LOW", High = "HIGH HIGH") or abnormal = list(Low = "LOW", High = "HIGH")
method (string) indicates how the returned map will be constructed. Can be "default" or "range".
na_str (string) string used to replace all NA or empty values in the output.

Value

A map data.frame.
Note

If method is "default", the returned map will only have the abnormal directions that are observed in the df, and records with all normal values will be excluded to avoid error in creating layout. If method is "range", the returned map will be based on the rule that at least one observation with low range > 0 for low direction and at least one observation with high range is not missing for high direction.

Examples

```r
adlb <- df_explicit_na(tern_ex_adlb)

h_map_for_count_abnormal(
  df = adlb,
  variables = list(anl = "ANRIND", split_rows = c("LBCAT", "PARAM")),
  abnormal = list(low = c("LOW"), high = c("HIGH")),
  method = "default",
  na_str = "<Missing>"
)

df <- data.frame(
  USUBJID = c(rep("1", 4), rep("2", 4), rep("3", 4)),
  AVISIT = c(
    rep("WEEK 1", 2),
    rep("WEEK 2", 2),
    rep("WEEK 1", 2),
    rep("WEEK 2", 2),
    rep("WEEK 1", 2),
    rep("WEEK 2", 2)
  ),
  PARAM = rep(c("ALT", "CPR"), 6),
  ANRIND = c(
    "NORMAL", "NORMAL", "LOW",
    "HIGH", "LOW", "LOW", "HIGH", "HIGH", rep("NORMAL", 4)
  ),
  ANRLO = rep(5, 12),
  ANRHI = rep(20, 12)
)
df$ANRIND <- factor(df$ANRIND, levels = c("LOW", "HIGH", "NORMAL"))

h_map_for_count_abnormal(
  df = df,
  variables = list(
    anl = "ANRIND",
    split_rows = c("PARAM"),
    range_low = "ANRLO",
    range_high = "ANRHI"
  ),
  abnormal = list(low = c("LOW"), high = c("HIGH")),
  method = "range",
  na_str = "<Missing>"
)
```

Helper functions for odds ratio estimation

Description

[Stable]
Functions to calculate odds ratios in `estimate_odds_ratio()`.

Usage

```r
or_glm(data, conf_level)
or_clogit(data, conf_level)
```

Arguments

data (data.frame): data frame containing at least the variables `rsp` and `grp`, and optionally `strata` for `or_clogit()`.
conf_level (proportion): confidence level of the interval.

Value

A named list of elements `or_ci` and `n_tot`.

Functions

- `or_glm()`: Estimates the odds ratio based on `stats::glm()`. Note that there must be exactly 2 groups in `data` as specified by the `grp` variable.
- `or_clogit()`: Estimates the odds ratio based on `survival::clogit()`. This is done for the whole data set including all groups, since the results are not the same as when doing pairwise comparisons between the groups.

See Also

`odds_ratio`

Examples

```r
# Data with 2 groups.
data <- data.frame(
  rsp = as.logical(c(1, 1, 0, 1, 0, 0, 1, 1)),
  grp = letters[c(1, 1, 1, 2, 2, 2, 1, 2)],
  strata = letters[c(1, 2, 1, 2, 2, 2, 1, 2)],
  stringsAsFactors = TRUE
)
```
# Odds ratio based on glm.
```
or_glm(data, conf_level = 0.95)
```

# Data with 3 groups.
```
data <- data.frame(  
  rsp = as.logical(c(1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1)),  
  grp = letters[c(1, 1, 1, 2, 2, 2, 3, 3, 3, 3, 1, 1, 1, 2, 2, 2, 3, 3, 3, 3)],  
  strata = LETTERS[c(1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2)],  
  stringsAsFactors = TRUE)
```

# Odds ratio based on stratified estimation by conditional logistic regression.
```
or_clogit(data, conf_level = 0.95)
```

---

**h_pkparam_sort**

Sort pharmacokinetic data by PARAM variable

---

**Description**

[Stable]

**Usage**

```
h_pkparam_sort(pk_data, key_var = "PARAMCD")
```

**Arguments**

- `pk_data` (data.frame): pharmacokinetic data frame.
- `key_var` (string): key variable used to merge pk_data and metadata created by `d_pkparam()`.

**Value**

A pharmacokinetic data.frame sorted by a PARAM variable.

**Examples**

```
library(dplyr)

adpp <- tern_ex_adpp %>% mutate(PKPARAM = factor(paste(PARAM, " (", AVALU, "),")))
pk_ordered_data <- h_pkparam_sort(adpp)
```
**h_proportions**

*Helper functions for calculating proportion confidence intervals*

---

**Description**

[Stable]

Functions to calculate different proportion confidence intervals for use in `estimate_proportion()`.

**Usage**

```r
prop_wilson(rsp, conf_level, correct = FALSE)

prop_strat_wilson(
  rsp,
  strata,
  weights = NULL,
  conf_level = 0.95,
  max_iterations = NULL,
  correct = FALSE
)

prop_clopper_pearson(rsp, conf_level)

prop_wald(rsp, conf_level, correct = FALSE)

prop_agresti_coull(rsp, conf_level)

prop_jeffreys(rsp, conf_level)
```

**Arguments**

- `rsp` *(logical)*
  - vector indicating whether each subject is a responder or not.

- `conf_level` *(proportion)*
  - confidence level of the interval.

- `correct` *(flag)*
  - whether to apply continuity correction.

- `strata` *(factor)*
  - variable with one level per stratum and same length as `rsp`.

- `weights` *(numeric or NULL)*
  - weights for each level of the strata. If `NULL`, they are estimated using the iterative algorithm proposed in Yan and Su (2010) that minimizes the weighted squared length of the confidence interval.

- `max_iterations` *(count)*
  - maximum number of iterations for the iterative procedure used to find estimates of optimal weights.
Value

Confidence interval of a proportion.

Functions

- `prop_wilson()`: Calculates the Wilson interval by calling `stats::prop.test()`. Also referred to as Wilson score interval.
- `prop_strat_wilson()`: Calculates the stratified Wilson confidence interval for unequal proportions as described in Yan and Su (2010)
- `prop_clopper_pearson()`: Calculates the Clopper-Pearson interval by calling `stats::binom.test()`. Also referred to as the exact method.
- `prop_wald()`: Calculates the Wald interval by following the usual textbook definition for a single proportion confidence interval using the normal approximation.
- `prop_agresti_coull()`: Calculates the Agresti-Coull interval. Constructed (for 95% CI) by adding two successes and two failures to the data and then using the Wald formula to construct a CI.
- `prop_jeffreys()`: Calculates the Jeffreys interval, an equal-tailed interval based on the non-informative Jeffreys prior for a binomial proportion.

References


See Also

`estimate_proportions`, descriptive function `d_proportion()`, and helper functions `strata_normal_quantile()` and `update_weights_strat_wilson()`.

Examples

```r
rsp <- c(
  TRUE, TRUE, TRUE, TRUE, TRUE,
  FALSE, FALSE, FALSE, FALSE, FALSE
)
prop_wilson(rsp, conf_level = 0.9)

# Stratified Wilson confidence interval with unequal probabilities

set.seed(1)
rsp <- sample(c(TRUE, FALSE), 100, TRUE)
strata_data <- data.frame(
  "f1" = sample(c("a", "b"), 100, TRUE),
  "f2" = sample(c("x", "y", "z"), 100, TRUE),
  stringsAsFactors = TRUE
)
strata <- interaction(strata_data)
n_strata <- ncol(table(rsp, strata)) # Number of strata
```
prop_strat_wilson(
    rsp = rsp, strata = strata,
    conf_level = 0.90
)

# Not automatic setting of weights
prop_strat_wilson(
    rsp = rsp, strata = strata,
    weights = rep(1 / n_strata, n_strata),
    conf_level = 0.90
)

prop_clopper_pearson(rsp, conf_level = .95)

prop_wald(rsp, conf_level = 0.95)
prop_wald(rsp, conf_level = 0.95, correct = TRUE)

prop_agresti_coull(rsp, conf_level = 0.95)

prop_jeffreys(rsp, conf_level = 0.95)

h_prop_diff

Helper functions to calculate proportion difference

Description

[Stable]

Usage

prop_diff_wald(rsp, grp, conf_level = 0.95, correct = FALSE)

prop_diff_ha(rsp, grp, conf_level)

prop_diff_nc(rsp, grp, conf_level, correct = FALSE)

prop_diff_cmh(rsp, grp, strata, conf_level = 0.95)

prop_diff_strat_nc(
    rsp,
    grp,
    strata,
    weights_method = c("cmh", "wilson_h"),
    conf_level = 0.95,
    correct = FALSE
)
Arguments

rsp (logical) vector indicating whether each subject is a responder or not.
grp (factor) vector assigning observations to one out of two groups (e.g. reference and treatment group).
conf_level (proportion) confidence level of the interval.
correct (flag) whether to include the continuity correction. For further information, see stats::prop.test().
strata (factor) variable with one level per stratum and same length as rsp.
weights_method (string) weights method. Can be either "cmh" or "heuristic" and directs the way weights are estimated.

Value

A named list of elements diff (proportion difference) and diff_ci (proportion difference confidence interval).

Functions

- prop_diff_wald(): The Wald interval follows the usual textbook definition for a single proportion confidence interval using the normal approximation. It is possible to include a continuity correction for Wald's interval.
- prop_diff_ha(): Anderson-Hauck confidence interval.
- prop_diff_nc(): Newcombe confidence interval. It is based on the Wilson score confidence interval for a single binomial proportion.
- prop_diff_cmh(): Calculates the weighted difference. This is defined as the difference in response rates between the experimental treatment group and the control treatment group, adjusted for stratification factors by applying Cochran-Mantel-Haenszel (CMH) weights. For the CMH chi-squared test, use stats::mantelhaen.test().
- prop_diff_strat_nc(): Calculates the stratified Newcombe confidence interval and difference in response rates between the experimental treatment group and the control treatment group, adjusted for stratification factors. This implementation follows closely the one proposed by Yan and Su (2010). Weights can be estimated from the heuristic proposed in prop_strat_wilson() or from CMH-derived weights (see prop_diff_cmh()).

References


See Also

prop_diff() for implementation of these helper functions.
Examples

# Wald confidence interval
set.seed(2)
rsp <- sample(c(TRUE, FALSE), replace = TRUE, size = 20)
grp <- factor(c(rep("A", 10), rep("B", 10)))

prop_diff_wald(rsp = rsp, grp = grp, conf_level = 0.95, correct = FALSE)

# Anderson-Hauck confidence interval
## "Mid" case: 3/4 respond in group A, 1/2 respond in group B.
rsp <- c(TRUE, FALSE, FALSE, TRUE, TRUE, TRUE)

prop_diff_ha(rsp = rsp, grp = grp, conf_level = 0.90)

## Edge case: Same proportion of response in A and B.
rsp <- c(TRUE, FALSE, TRUE, FALSE)
grp <- factor(c("A", "A", "B", "B"), levels = c("A", "B"))

prop_diff_ha(rsp = rsp, grp = grp, conf_level = 0.6)

# Newcombe confidence interval
set.seed(1)
rsp <- c(  
sample(c(TRUE, FALSE), size = 40, prob = c(3 / 4, 1 / 4), replace = TRUE),  
sample(c(TRUE, FALSE), size = 40, prob = c(1 / 2, 1 / 2), replace = TRUE)
)
grp <- factor(rep(c("A", "B"), each = 40), levels = c("B", "A"))
table(rsp, grp)

prop_diff_nc(rsp = rsp, grp = grp, conf_level = 0.9)

# Cochran-Mantel-Haenszel confidence interval
set.seed(2)
rsp <- sample(c(TRUE, FALSE), 100, TRUE)
grp <- sample(c("Placebo", "Treatment"), 100, TRUE)
grp <- factor(grp, levels = c("Placebo", "Treatment"))
strata_data <- data.frame(  
  "f1" = sample(c("a", "b"), 100, TRUE),  
  "f2" = sample(c("x", "y", "z"), 100, TRUE),  
  stringsAsFactors = TRUE
)

prop_diff_cmh(  
rsp = rsp, grp = grp, strata = interaction(strata_data),  
conf_level = 0.90
)

# Stratified Newcombe confidence interval
```r
set.seed(2)
data_set <- data.frame(
  "rsp" = sample(c(TRUE, FALSE), 100, TRUE),
  "f1" = sample(c("a", "b"), 100, TRUE),
  "f2" = sample(c("x", "y", "z"), 100, TRUE),
  "grp" = sample(c("Placebo", "Treatment"), 100, TRUE),
  stringsAsFactors = TRUE
)

prop_diff_strat_nc(
  rsp = data_set$rsp, grp = data_set$grp, strata = interaction(data_set[2:3]),
  weights_method = "cmh",
  conf_level = 0.90
)

prop_diff_strat_nc(
  rsp = data_set$rsp, grp = data_set$grp, strata = interaction(data_set[2:3]),
  weights_method = "wilson_h",
  conf_level = 0.90
)
```

---

### h_response_biomarkers_subgroups

*Helper functions for tabulating biomarker effects on binary response by subgroup*

#### Description

*Stable*

Helper functions which are documented here separately to not confuse the user when reading about the user-facing functions.

#### Usage

- `h_rsp_to_logistic_variables(variables, biomarker)`
- `h_logistic_mult_cont_df(variables, data, control = control_logistic())`
- `h_tab_rsp_one_biomarker(df, vars, na_str = default_na_str(), .indent.mods = 0L)`

#### Arguments

- `variables` *(named list of string)*
  list of additional analysis variables.
- `biomarker` *(string)*
  the name of the biomarker variable.
h_response_biomarkers_subgroups

data (data.frame) the dataset containing the variables to summarize.

control (named list) controls for the response definition and the confidence level produced by control_logistic().

df (data.frame) results for a single biomarker, as part of what is returned by extract_rsp_biomarkers() (it needs a couple of columns which are added by that high-level function relative to what is returned by h_logistic_mult_cont_df(), see the example).

vars (character) the names of statistics to be reported among:
  • n_tot: Total number of patients per group.
  • n_rsp: Total number of responses per group.
  • prop: Total response proportion per group.
  • or: Odds ratio.
  • ci: Confidence interval of odds ratio.
  • pval: p-value of the effect. Note, the statistics n_tot, or and ci are required.

na_str (string) string used to replace all NA or empty values in the output.

.indent.mods (named integer) indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

Value

• h_rsp_to_logistic_variables() returns a named list of elements response, arm, covariates, and strata.

• h_logistic_mult_cont_df() returns a data.frame containing estimates and statistics for the selected biomarkers.

• h_tab_rsp_one_biomarker() returns an rtables table object with the given statistics arranged in columns.

Functions

• h_rsp_to_logistic_variables(): helps with converting the "response" function variable list to the "logistic regression" variable list. The reason is that currently there is an inconsistency between the variable names accepted by extract_rsp_subgroups() and fit_logistic().

• h_logistic_mult_cont_df(): prepares estimates for number of responses, patients and overall response rate, as well as odds ratio estimates, confidence intervals and p-values, for multiple biomarkers in a given single data set. variables corresponds to names of variables found in data, passed as a named list and requires elements rsp and biomarkers (vector of continuous biomarker variables) and optionally covariates and strata.

• h_tab_rsp_one_biomarker(): Prepares a single sub-table given a df_sub containing the results for a single biomarker.
Examples

```r
library(dplyr)
library(forcats)

adrs <- tern_ex_adrs
adrs_labels <- formatters::var_labels(adrs)

adrs_f <- adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  mutate(rsp = AVALC == "CR")
formatters::var_labels(adrs_f) <- c(adrs_labels, "Response")

# This is how the variable list is converted internally.

h_rsp_to_logistic_variables(
  variables = list(
    rsp = "RSP",
    covariates = c("A", "B"),
    strata = "D",
  ),
  biomarker = "AGE"
)

# For a single population, estimate separately the effects
# of two biomarkers.

df <- h_logistic_mult_cont_df(
  variables = list(
    rsp = "rsp",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
  ),
  data = adrs_f
)

df

# If the data set is empty, still the corresponding rows with missings are returned.

h_coxreg_mult_cont_df(
  variables = list(
    rsp = "rsp",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
    strata = "STRATA1"
  ),
  data = adrs_f[NULL, ]
)

# Starting from above 'df', zoom in on one biomarker and add required columns.

dfl <- df[1, ]
dfl$subgroup <- "All patients"
dfl$row_type <- "content"
dfl$var <- "ALL"
dfl$var_label <- "All patients"
```
h_response_subgroups

Helper functions for tabulating binary response by subgroup

Description

[Stable]
Helper functions that tabulate in a data frame statistics such as response rate and odds ratio for population subgroups.

Usage

h_proportion_df(rsp, arm)

h_proportion_subgroups_df(
    variables,
    data,
    groups_lists = list(),
    label_all = "All Patients"
)

h_odds_ratio_df(rsp, arm, strata_data = NULL, conf_level = 0.95, method = NULL)

h_odds_ratio_subgroups_df(
    variables,
    data,
    groups_lists = list(),
    conf_level = 0.95,
    method = NULL,
    label_all = "All Patients"
)

Arguments

rsp (logical)
vector indicating whether each subject is a responder or not.

arm (factor)
the treatment group variable.

variables (named list of string)
list of additional analysis variables.

data (data.frame)
the dataset containing the variables to summarize.
groups_lists   (named list of list)
optionally contains for each subgroups variable a list, which specifies the new
group levels via the names and the levels that belong to it in the character vectors
that are elements of the list.

label_all   (string)
label for the total population analysis.

strata_data   (factor, data.frame, or NULL)
required if stratified analysis is performed.

conf_level   (proportion)
confidence level of the interval.

method   (string or NULL)
specifies the test used to calculate the p-value for the difference between two
proportions. For options, see s_test_proportion_diff(). Default is NULL so
no test is performed.

Details

Main functionality is to prepare data for use in a layout-creating function.

Value

•  h_proportion_df() returns a data.frame with columns arm, n, n_rsp, and prop.

•  h_proportion_subgroups_df() returns a data.frame with columns arm, n, n_rsp, prop,
   subgroup, var, var_label, and row_type.

•  h_odds_ratio_df() returns a data.frame with columns arm, n_tot, or, lcl, ucl, conf_level,
   and optionally pval and pval_label.

•  h_odds_ratio_subgroups_df() returns a data.frame with columns arm, n_tot, or, lcl,
   ucl, conf_level, subgroup, var, var_label, and row_type.

Functions

•  h_proportion_df(): Helper to prepare a data frame of binary responses by arm.

•  h_proportion_subgroups_df(): Summarizes proportion of binary responses by arm and
   across subgroups in a data frame. variables corresponds to the names of variables found
   in data, passed as a named list and requires elements rsp, arm and optionally subgroups.
groups_lists optionally specifies groupings for subgroups variables.

•  h_odds_ratio_df(): Helper to prepare a data frame with estimates of the odds ratio between
   a treatment and a control arm.

•  h_odds_ratio_subgroups_df(): Summarizes estimates of the odds ratio between a treat-
   ment and a control arm across subgroups in a data frame. variables corresponds to the names
   of variables found in data, passed as a named list and requires elements rsp, arm and option-
   ally subgroups and strata. groups_lists optionally specifies groupings for subgroups
   variables.
library(dplyr)
library(forcats)

adrs <- tern_ex_adrs
adrs_labels <- formatters::var_labels(adrs)

adrs_f <- adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(ARM %in% c("A: Drug X", "B: Placebo")) %>%
  droplevels() %>%
  mutate(
    # Reorder levels of factor to make the placebo group the reference arm.
    ARM = fct_relevel(ARM, "B: Placebo"),
    rsp = AVALC == "CR"
  )
formatters::var_labels(adrs_f) <- c(adrs_labels, "Response")

h_proportion_df(c(TRUE, FALSE, FALSE), arm = factor(c("A", "A", "B"), levels = c("A", "B")))

h_proportion_subgroups_df(
  variables = list(rsp = "rsp", arm = "ARM", subgroups = c("SEX", "BMRKR2")),
  data = adrs_f
)

# Define groupings for BMRKR2 levels.

h_proportion_subgroups_df(
  variables = list(rsp = "rsp", arm = "ARM", subgroups = c("SEX", "BMRKR2")),
  data = adrs_f,
  groups_lists = list(
    BMRKR2 = list(
      "low" = "LOW",
      "low/medium" = c("LOW", "MEDIUM"),
      "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
    )
  )
)

# Unstratified analysis.

h_odds_ratio_df(c(TRUE, FALSE, FALSE, TRUE), arm = factor(c("A", "A", "B", "B"), levels = c("A", "B")))

# Include p-value.

h_odds_ratio_df(adrs_f$rsp, adrs_f$ARM, method = "chisq")

# Stratified analysis.

h_odds_ratio_df(100x712)e

"h_response_subgroups"

Examples

library(dplyr)
library(forcats)

adrs <- tern_ex_adrs
adrs_labels <- formatters::var_labels(adrs)

adrs_f <- adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(ARM %in% c("A: Drug X", "B: Placebo")) %>%
  droplevels() %>%
  mutate(
    # Reorder levels of factor to make the placebo group the reference arm.
    ARM = fct_relevel(ARM, "B: Placebo"),
    rsp = AVALC == "CR"
  )
formatters::var_labels(adrs_f) <- c(adrs_labels, "Response")

h_proportion_df(c(TRUE, FALSE, FALSE), arm = factor(c("A", "A", "B"), levels = c("A", "B")))

h_proportion_subgroups_df(
  variables = list(rsp = "rsp", arm = "ARM", subgroups = c("SEX", "BMRKR2")),
  data = adrs_f
)

# Define groupings for BMRKR2 levels.

h_proportion_subgroups_df(
  variables = list(rsp = "rsp", arm = "ARM", subgroups = c("SEX", "BMRKR2")),
  data = adrs_f,
  groups_lists = list(
    BMRKR2 = list(
      "low" = "LOW",
      "low/medium" = c("LOW", "MEDIUM"),
      "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
    )
  )
)

# Unstratified analysis.

h_odds_ratio_df(c(TRUE, FALSE, FALSE, TRUE), arm = factor(c("A", "A", "B", "B"), levels = c("A", "B")))

# Include p-value.

h_odds_ratio_df(adrs_f$rsp, adrs_f$ARM, method = "chisq")

# Stratified analysis.

h_odds_ratio_df(100x712)e
h_split_by_subgroups

Split data frame by subgroups

Description

[Stable]

Split a data frame into a non-nested list of subsets.
Usage

```r
h_split_by_subgroups(data, subgroups, groups_lists = list())
```

Arguments

- `data` (data.frame): dataset to split.
- `subgroups` (character): names of factor variables from `data` used to create subsets. Unused levels not present in `data` are dropped. Note that the order in this vector determines the order in the downstream table.
- `groups_lists` (named list of list): optionally contains for each `subgroups` variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.

Details

Main functionality is to prepare data for use in forest plot layouts.

Value

A list with subset data (`df`) and metadata about the subset (`df_labels`).

Examples

```r
df <- data.frame(
  x = c(1:5),
  y = factor(c("A", "B", "A", "B", "A"), levels = c("A", "B", "C")),
  z = factor(c("C", "C", "D", "D", "D"), levels = c("D", "C"))
)
formatters::var_labels(df) <- paste("label for", names(df))

h_split_by_subgroups(
  data = df,
  subgroups = c("y", "z")
)

h_split_by_subgroups(
  data = df,
  subgroups = c("y", "z"),
  groups_lists = list(
    y = list("AB" = c("A", "B"), "C" = "C")
  )
)
```
h_split_param

Split parameters

Description

[Stable]

It divides the data in the vector `param` into the groups defined by `f` based on specified values. It is relevant in `rtables` layers so as to distribute parameters `.stats` or `.formats` into lists with items corresponding to specific analysis function.

Usage

```r
h_split_param(param, value, f)
```

Arguments

- `param` (vector): the parameter to be split.
- `value` (vector): the value used to split.
- `f` (list): the reference to make the split.

Value

A named list with the same element names as `f`, each containing the elements specified in `.stats`.

Examples

```r
f <- list(
  surv = c("pt_at_risk", "event_free_rate", "rate_se", "rate_ci"),
  surv_diff = c("rate_diff", "rate_diff_ci", "ztest_pval")
)

.stats <- c("pt_at_risk", "rate_diff")
h_split_param(.stats, .stats, f = f)
# $surv
# [1] "pt_at_risk"
#
# $surv_diff
# [1] "rate_diff"

.formats <- c("pt_at_risk" = "xx", "event_free_rate" = "xxx")
h_split_param(.formats, names(.formats), f = f)
# $surv
# pt_at_risk event_free_rate
```
h_stack_by_baskets  

Helper function to create a new SMQ variable in ADAE by stacking SMQ and/or CQ records.

Description

[Stable]

Helper function to create a new SMQ variable in ADAE that consists of all adverse events belonging to selected Standardized/Customized queries. The new dataset will only contain records of the adverse events belonging to any of the selected baskets. Remember that na_str must match the needed pre-processing done with df_explicit_na() to have the desired output.

Usage

```r
h_stack_by_baskets(
  df,
  baskets = grep("^(SMQ|CQ).+NAM$", names(df), value = TRUE),
  smq_varlabel = "Standardized MedDRA Query",
  keys = c("STUDYID", "USUBJID", "ASTDTM", "AEDECOD", "AESEQ"),
  aag_summary = NULL,
  na_str = "<Missing>"
)
```

Arguments

df  
(data.frame)  
data set containing all analysis variables.

baskets  
(character)  
variable names of the selected Standardized/Customized queries.

smq_varlabel  
(string)  
a label for the new variable created.

keys  
(character)  
names of the key variables to be returned along with the new variable created.

aag_summary  
(data.frame)  
containing the SMQ baskets and the levels of interest for the final SMQ variable. This is useful when there are some levels of interest that are not observed in the df dataset. The two columns of this dataset should be named basket and basket_name.

na_str  
(string)  
string used to replace all NA or empty values in the output.
Value

A data.frame with variables in keys taken from df and new variable SMQ containing records belonging to the baskets selected via the baskets argument.

Examples

```r
da <- tern_ex_ae[1:20, ] %>% df_explicit_na()
h_stack_by_baskets(df = dae)

aag <- data.frame(
  NAMVAR = c("CQ01NAM", "CQ02NAM", "SMQ01NAM", "SMQ02NAM"),
  REFNAME = c(
    "D.2.1.5.3/A.1.1.1.1 aesi", "X.9.9.9.9/Y.8.8.8.8 aesi",
    "C.1.1.1.3/B.2.2.3.1 aesi", "C.1.1.1.3/B.3.3.3.3 aesi" ),
  SCOPE = c("", ",", "BROAD", "BROAD"),
  stringsAsFactors = FALSE
)

basket_name <- character(nrow(aag))
cq_pos <- grep("^CQ.*NAM$", aag$NAMVAR)
smq_pos <- grep("^SMQ.*NAM$", aag$NAMVAR)
basket_name[cq_pos] <- aag$REFNAME[cq_pos]
basket_name[smq_pos] <- paste0(
  aag$REFNAME[smq_pos], ",", aag$SCOPE[smq_pos], ",")

aag_summary <- data.frame(
  basket = aag$NAMVAR,
  basket_name = basket_name,
  stringsAsFactors = TRUE
)

result <- h_stack_by_baskets(df = dae, aag_summary = aag_summary)
all(levels(aag_summary$basket_name) %in% levels(result$SMQ))
```

h_step

Helper functions for subgroup treatment effect pattern (STEP) calculations
Description

[Stable]
Helper functions that are used internally for the STEP calculations.

Usage

h_step_window(x, control = control_step())

h_step_trt_effect(data, model, variables, x)

h_step_survival_formula(variables, control = control_step())

h_step_survival_est(
  formula,
  data,
  variables,
  x,
  subset = rep(TRUE, nrow(data)),
  control = control_coxph()
)

h_step_rsp_formula(variables, control = c(control_step(), control_logistic()))

h_step_rsp_est(
  formula,
  data,
  variables,
  x,
  subset = rep(TRUE, nrow(data)),
  control = control_logistic()
)

Arguments

x (numeric)
  biomarker value(s) to use (without NA).

control (named list)
  output from control_step().

data (data.frame)
  the dataset containing the variables to summarize.

model (coxph or glm)
  the regression model object.

variables (named list of string)
  list of additional analysis variables.

formula (formula)
  the regression model formula.
subset (logical)
subset vector.

Value

- h_step_window() returns a list containing the window-selection matrix sel and the interval information matrix interval.
- h_step_trt_effect() returns a vector with elements est and se.
- h_step_survival_formula() returns a model formula.
- h_step_survival_est() returns a matrix of number of observations n, events, log hazard ratio estimates loghr, standard error se, and Wald confidence interval bounds ci_lower and ci_upper. One row is included for each biomarker value in x.
- h_step_rsp_formula() returns a model formula.
- h_step_rsp_est() returns a matrix of number of observations n, log odds ratio estimates logor, standard error se, and Wald confidence interval bounds ci_lower and ci_upper. One row is included for each biomarker value in x.

Functions

- h_step_window(): Creates the windows for STEP, based on the control settings provided.
- h_step_trt_effect(): Calculates the estimated treatment effect estimate on the linear predictor scale and corresponding standard error from a STEP model fitted on data given variables specification, for a single biomarker value x. This works for both coxph and glm models, i.e. for calculating log hazard ratio or log odds ratio estimates.
- h_step_survival_formula(): Builds the model formula used in survival STEP calculations.
- h_step_survival_est(): Estimates the model with formula built based on variables in data for a given subset and control parameters for the Cox regression.
- h_step_rsp_formula(): Builds the model formula used in response STEP calculations.
- h_step_rsp_est(): Estimates the model with formula built based on variables in data for a given subset and control parameters for the logistic regression.
h_surv_to_coxreg_variables(variables, biomarker)

h_coxreg_mult_cont_df(variables, data, control = control_coxreg())

h_tab_surv_one_biomarker(
  df,
  vars,
  time_unit,
  na_str = default_na_str(),
  .indent_mods = 0L,
  ...
)

Arguments

variables (named list of string)
  list of additional analysis variables.

biomarker (string)
  the name of the biomarker variable.

data (data.frame)
  the dataset containing the variables to summarize.

control (list)
  a list of parameters as returned by the helper function control_coxreg().

df (data.frame)
  results for a single biomarker, as part of what is returned by extract_survival_biomarkers()
  (it needs a couple of columns which are added by that high-level function relative to what is returned by h_coxreg_mult_cont_df(), see the example).

vars (character)
  the names of statistics to be reported among:
  • n_tot_events: Total number of events per group.
  • n_tot: Total number of observations per group.
  • median: Median survival time.
  • hr: Hazard ratio.
  • ci: Confidence interval of hazard ratio.
  • pval: p-value of the effect. Note, one of the statistics n_tot and n_tot_events, as well as both hr and ci are required.

time_unit (string)
  label with unit of median survival time. Default NULL skips displaying unit.

na_str (string)
  string used to replace all NA or empty values in the output.

.indent_mods (named integer)
  indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

... additional arguments for the lower level functions.
**Value**

- `h_surv_to_coxreg_variables()` returns a named list of elements `time`, `event`, `arm`, `covariates`, and `strata`.

- `h_coxreg_mult_cont_df()` returns a `data.frame` containing estimates and statistics for the selected biomarkers.

- `h_tab_surv_one_biomarker()` returns an `rtables` table object with the given statistics arranged in columns.

**Functions**

- `h_surv_to_coxreg_variables()`: Helps with converting the "survival" function variable list to the "Cox regression" variable list. The reason is that currently there is an inconsistency between the variable names accepted by `extract_survival_subgroups()` and `fit_coxreg_multivar()`.

- `h_coxreg_mult_cont_df()`: Prepares estimates for number of events, patients and median survival times, as well as hazard ratio estimates, confidence intervals and p-values, for multiple biomarkers in a given single data set. `variables` corresponds to names of variables found in `data`, passed as a named list and requires elements `tte`, `is_event`, `biomarkers` (vector of continuous biomarker variables) and optionally `subgroups` and `strata`.

- `h_tab_surv_one_biomarker()`: Prepares a single sub-table given a `df_sub` containing the results for a single biomarker.

**Examples**

```r
library(dplyr)
library(forcats)

adtte <- tern_ex_adtte

# Save variable labels before data processing steps.
adtte_labels <- formatters::var_labels(adtte, fill = FALSE)

adtte_f <- adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(
    AVALU = as.character(AVALU),
    is_event = CNSR == 0
  )

labels <- c("AVALU" = adtte_labels[["AVALU"]], "is_event" = "Event Flag")
formatters::var_labels(adtte_f)[names(labels)] <- labels

# This is how the variable list is converted internally.
h_surv_to_coxreg_variables(
  variables = list(
    tte = "AVAL",
    is_event = "EVNT",
    covariates = c("A", "B"),
    strata = "D"
  ),
)```
biomarker = "AGE"
)

# For a single population, estimate separately the effects
# of two biomarkers.
df <- h_coxreg_mult_cont_df(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
    strata = c("STRATA1", "STRATA2")
  ),
  data = adtte_f
)
df

# If the data set is empty, still the corresponding rows with missings are returned.
h_coxreg_mult_cont_df(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "REGION1",
    strata = c("STRATA1", "STRATA2")
  ),
  data = adtte_f[NULL, ]
)

# Starting from above `df`, zoom in on one biomarker and add required columns.
df1 <- df[1, ]
df1$subgroup <- "All patients"
df1$row_type <- "content"
df1$var <- "ALL"
df1$var_label <- "All patients"
h_tab_surv_one_biomarker(
  df1,
  vars = c("n_tot", "n_tot_events", "median", "hr", "ci", "pval"),
  time_unit = "days"
)

---

**h_survival_duration_subgroups**

*Helper functions for tabulating survival duration by subgroup*

**Description**

*[Stable]*

Helper functions that tabulate in a data frame statistics such as median survival time and hazard ratio for population subgroups.
Usage

h_survtime_df(tte, is_event, arm)

h_survtime_subgroups_df(
  variables,
  data,
  groups_lists = list(),
  label_all = "All Patients"
)

h_coxph_df(tte, is_event, arm, strata_data = NULL, control = control_coxph())

h_coxph_subgroups_df(
  variables,
  data,
  groups_lists = list(),
  control = control_coxph(),
  label_all = "All Patients"
)

Arguments

tte (numeric)
  vector of time-to-event duration values.

is_event (flag)
  TRUE if event, FALSE if time to event is censored.

arm (factor)
  the treatment group variable.

variables (named list of string)
  list of additional analysis variables.

data (data.frame)
  the dataset containing the variables to summarize.

groups_lists (named list of list)
  optionally contains for each subgroups variable a list, which specifies the new
  group levels via the names and the levels that belong to it in the character vectors
  that are elements of the list.

label_all (string)
  label for the total population analysis.

strata_data (factor, data.frame, or NULL)
  required if stratified analysis is performed.

ccontrol (list)
  parameters for comparison details, specified by using the helper function control_coxph().
  Some possible parameter options are:
  • pval_method (string)
    p-value method for testing hazard ratio = 1. Default method is "log-rank"
which comes from `survival::survdiff()`, can also be set to "wald" or "likelihood" (from `survival::coxph()`).

- **ties** (string)
  specifying the method for tie handling. Default is "efron", can also be set to "breslow" or "exact". See more in `survival::coxph()`

- **conf_level** (proportion)
  confidence level of the interval for HR.

### Details
Main functionality is to prepare data for use in a layout-creating function.

### Value
- `h_survtime_df()` returns a data.frame with columns `arm`, `n`, `n_events`, and `median`.
- `h_survtime_subgroups_df()` returns a data.frame with columns `arm`, `n`, `n_events`, `median`, `subgroup`, `var`, `var_label`, and `row_type`.
- `h_coxph_df()` returns a data.frame with columns `arm`, `n_tot`, `n_tot_events`, `hr`, `lcl`, `ucl`, `conf_level`, `pval` and `pval_label`.
- `h_coxph_subgroups_df()` returns a data.frame with columns `arm`, `n_tot`, `n_tot_events`, `hr`, `lcl`, `ucl`, `conf_level`, `pval`, `pval_label`, `subgroup`, `var`, `var_label`, and `row_type`.

### Functions
- `h_survtime_df()`: Helper to prepare a data frame of median survival times by arm.
- `h_survtime_subgroups_df()`: Summarizes median survival times by arm and across subgroups in a data frame. variables corresponds to the names of variables found in data, passed as a named list and requires elements `tte`, `is_event`, `arm` and optionally `subgroups`. `groups_lists` optionally specifies groupings for subgroups variables.
- `h_coxph_df()`: Helper to prepare a data frame with estimates of treatment hazard ratio.
- `h_coxph_subgroups_df()`: Summarizes estimates of the treatment hazard ratio across subgroups in a data frame. variables corresponds to the names of variables found in data, passed as a named list and requires elements `tte`, `is_event`, `arm` and optionally `subgroups` and `strata`. `groups_lists` optionally specifies groupings for subgroups variables.

### Examples
```r
library(dplyr)
library(forcats)

adtte <- tern_ex_adtte
# Save variable labels before data processing steps.
adtte_labels <- formatters::var_labels(adtte)
adtte_f <- adtte %>%
  filter(}
```
PARAMCD == "OS",
ARM %in% c("B: Placebo", "A: Drug X"),
SEX %in% c("M", "F")
)

mutate(
  # Reorder levels of ARM to display reference arm before treatment arm.
  ARM = droplevels(fct_relevel(ARM, "B: Placebo")),
  SEX = droplevels(SEX),
  is_event = CNSR == 0
)

labels <- c("ARM" = adtte_labels["ARM"], "SEX" = adtte_labels["SEX"], "is_event" = "Event Flag")
formatters::var_labels(adtte_f)[names(labels)] <- labels

# Extract median survival time for one group.

h_survtime_df(
  tte = adtte_f$AVAL,
  is_event = adtte_f$is_event,
  arm = adtte_f$ARM
)

# Extract median survival time for multiple groups.

h_survtime_subgroups_df(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    arm = "ARM",
    subgroups = c("SEX", "BMRKR2")
  ),
  data = adtte_f
)

# Define groupings for BMRKR2 levels.

h_survtime_subgroups_df(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    arm = "ARM",
    subgroups = c("SEX", "BMRKR2")
  ),
  data = adtte_f,
  groups_lists = list(
    BMRKR2 = list(
      "low" = "LOW",
      "low/medium" = c("LOW", "MEDIUM"),
      "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
    )
  )
)

# Extract hazard ratio for one group.

h_coxph_df(adtte_f$AVAL, adtte_f$is_event, adtte_f$ARM)

# Extract hazard ratio for one group with stratification factor.
h_coxph_df(adtte_f$AVAL, adtte_f$is_event, adtte_f$ARM, strata_data = adtte_f$STRATA1)

# Extract hazard ratio for multiple groups.
h_coxph_subgroups_df(
    variables = list(
        tte = "AVAL",
        is_event = "is_event",
        arm = "ARM",
        subgroups = c("SEX", "BMRKR2")
    ),
    data = adtte_f
)

# Define groupings of BMRKR2 levels.
h_coxph_subgroups_df(
    variables = list(
        tte = "AVAL",
        is_event = "is_event",
        arm = "ARM",
        subgroups = c("SEX", "BMRKR2")
    ),
    data = adtte_f,
    groups_lists = list(
        BMRKR2 = list(
            "low" = "LOW",
            "low/medium" = c("LOW", "MEDIUM"),
            "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
        )
    )
)

# Extract hazard ratio for multiple groups with stratification factors.
h_coxph_subgroups_df(
    variables = list(
        tte = "AVAL",
        is_event = "is_event",
        arm = "ARM",
        subgroups = c("SEX", "BMRKR2"),
        strata = c("STRATA1", "STRATA2")
    ),
    data = adtte_f
)

---

h_tab_one_biomarker  

**Helper function for tabulation of a single biomarker result**

**Description**

[Stable]
Please see `h_tab_surv_one_biomarker()` and `h_tab_rsp_one_biomarker()`, which use this function for examples. This function is a wrapper for `rtables::summarize_row_groups()`.

Usage

```r
def
afuns
colvars
na_str = default_na_str(),
indent_mods = 0L,
...
)
```

Arguments

df (data.frame) results for a single biomarker.

afuns (named list of function) analysis functions.

colvars (named list) named list with elements vars (variables to tabulate) and labels (their labels).

na_str (string) string used to replace all NA or empty values in the output.

.indent_mods (named integer) indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

... additional arguments for the lower level functions.

Value

An rtables table object with statistics in columns.

---

'h_tbl_coxph_pairwise' 'Helper function for generating a pairwise Cox-PH table'

Description

[Stable]
Create a data.frame of pairwise stratified or unstratified Cox-PH analysis results.
Usage

```r
h_tbl_coxph_pairwise(
  df, 
  variables, 
  ref_group_coxph = NULL, 
  control_coxph_pw = control_coxph(), 
  annot_coxph_refLbls = FALSE 
)
```

Arguments

df (data.frame) data set containing all analysis variables.

variables (named list) variable names. Details are:
  • tte (numeric) variable indicating time-to-event duration values.
  • is_event (logical) event variable. TRUE if event, FALSE if time to event is censored.
  • arm (factor) the treatment group variable.
  • strata (character or NULL) variable names indicating stratification factors.

ref_group_coxph (string or NULL) level of arm variable to use as reference group in calculations for `annot_coxph` table. If NULL (default), uses the first level of the arm variable.

control_coxph_pw (list) parameters for comparison details, specified using the helper function `control_coxph()`. Some possible parameter options are:
  • pval_method (string) p-value method for testing hazard ratio = 1. Default method is "log-rank", can also be set to "wald" or "likelihood".
  • ties (string) method for tie handling. Default is "efron", can also be set to "breslow" or "exact". See more in `survival::coxph()`
  • conf_level (proportion) confidence level of the interval for HR.

annot_coxph_refLbls (flag) whether the reference group should be explicitly printed in labels for the `annot_coxph` table. If FALSE (default), only comparison groups will be printed in `annot_coxph` table labels.
A data.frame containing statistics HR, XX% CI (XX taken from control_coxph_pw), and p-value (log-rank).

Examples

```r
library(dplyr)

adtte <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(is_event = CNSR == 0)

h_tbl_coxph_pairwise(
  df = adtte,
  variables = list(tte = "AVAL", is_event = "is_event", arm = "ARM"),
  control_coxph_pw = control_coxph(conf_level = 0.9)
)
```

Description

[Stable]

Transform a survival fit to a table with groups in rows characterized by N, median and confidence interval.

Usage

```r
h_tbl_median_surv(fit_km, armval = "All")
```

Arguments

- `fit_km` (survfit): result of `survival::survfit()`.
- `armval` (string): used as strata name when treatment arm variable only has one level. Default is "All".

Value

A summary table with statistics N, Median, and XX% CI (XX taken from fit_km).
Examples

```r
library(dplyr)
library(survival)

adtte <- tern_ex_adtte %>% filter(PARAMCD == "OS")
fit <- survfit(
    formula = Surv(AVAL, 1 - CNSR) ~ ARMCD,
    data = adtte
)
h_tbl_median_surv(fit_km = fit)
```

---

```r
def h_worsen_counter(df, id, .var, baseline_var, direction_var)

Arguments

df (data.frame)
data set containing all analysis variables.

id (string)
subject variable name.

.var (string)
single variable name that is passed by rtables when requested by a statistics function.

baseline_var (string)
name of the baseline lab grade variable.

direction_var (string)
name of the direction variable specifying the direction of the shift table of interest. Only lab records flagged by L, H or B are included in the shift table.

- L: low direction only
- H: high direction only
- B: both low and high directions
```

**h_worsen_counter**

Helper function to analyze patients for `s_count_abnormal_lab_worsen_by_baseline()`.

**Description**

[Stable]

Helper function to count the number of patients and the fraction of patients according to highest post-baseline lab grade variable `.var`, baseline lab grade variable `baseline_var`, and the direction of interest specified in `direction_var`. 

**Usage**

```r
h_worsen_counter(df, id, .var, baseline_var, direction_var)
```
Value

The counts and fraction of patients whose worst post-baseline lab grades are worse than their baseline grades, for post-baseline worst grades "1", "2", "3", "4" and "Any".

See Also

abnormal_by_worst_grade_worsen

Examples

library(dplyr)

# The direction variable, GRADDR, is based on metadata
adlb <- tern_ex_adlb %>%
  mutate(
    GRADDR = case_when(
      PARAMCD == "ALT" ~ "B",
      PARAMCD == "CRP" ~ "L",
      PARAMCD == "IGA" ~ "H"
    )
  ) %>%
  filter(SAFFL == "Y" & ONTRTFL == "Y" & GRADDR != "")

df <- h_adlb_worsen(
adlb,
  worst_flag_low = c("WGRLOFL" = "Y"),
  worst_flag_high = c("WGRHIFL" = "Y"),
  direction_var = "GRADDR"
)

# `h_worsen_counter`

h_worsen_counter(  
df %>% filter(PARAMCD == "CRP" & GRADDR == "Low"),
  id = "USUBJID",
  .var = "ATOXGR",
  baseline_var = "BTOXGR",
  direction_var = "GRADDR"
)

__h_xticks__

**Helper function to calculate x-tick positions**

Description

[Stable]

Calculate the positions of ticks on the x-axis. However, if xticks already exists it is kept as is. It is based on the same function ggplot2 relies on, and is required in the graphic and the patient-at-risk annotation table.
Usage

h_xticks(data, xticks = NULL, max_time = NULL)

Arguments

data (data.frame) survival data as pre-processed by h_data_plot.
xticks (numeric or NULL) numeric vector of tick positions or a single number with spacing between ticks on the x-axis. If NULL (default), labeling::extended() is used to determine optimal tick positions on the x-axis.
max_time (numeric(1)) maximum value to show on x-axis. Only data values less than or up to this threshold value will be plotted (defaults to NULL).

Value

A vector of positions to use for x-axis ticks on a ggplot object.

Examples

library(dplyr)
library(survival)

data <- tern_ex_adtte %>% filter(PARAMCD == "OS") %>% survfit(formula = Surv(AVAL, 1 - CNSR) ~ ARMCD, data = .) %>% h_data_plot()

h_xticks(data)
h_xticks(data, xticks = seq(0, 3000, 500))
h_xticks(data, xticks = 500)
h_xticks(data, xticks = 500, max_time = 6000)
h_xticks(data, xticks = c(0, 500), max_time = 300)
h_xticks(data, xticks = 500, max_time = 300)
imputation_rule

Usage

imputation_rule(
  df,
  x_stats,
  stat,
  imp_rule,
  post = FALSE,
  avalcat_var = "AVALCAT1"
)

Arguments

df      (data.frame) data set containing all analysis variables.

x_stats (named list) a named list of statistics, typically the results of s_summary().

stat    (string) statistic to return the value/NA level of according to the imputation rule applied.

imp_rule (string) imputation rule setting. Set to "1/3" to implement 1/3 imputation rule or "1/2" to implement 1/2 imputation rule.

post    (flag) whether the data corresponds to a post-dose time-point (defaults to FALSE). This parameter is only used when imp_rule is set to "1/3".

avalcat_var (string) name of variable that indicates whether a row in df corresponds to an analysis value in category "BLQ", "LTR", "<PCLLOQ", or none of the above (defaults to "AVALCAT1"). Variable avalcat_var must be present in df.

Value

A list containing statistic value (val) and NA level (na_str) that should be displayed according to the specified imputation rule.

See Also

analyze_vars_in_cols() where this function can be implemented by setting the imp_rule argument.

Examples

set.seed(1)
df <- data.frame(
  AVAL = runif(50, 0, 1),
  AVALCAT1 = sample(c(1, "BLQ"), 50, replace = TRUE)
)
x_stats <- s_summary(df$AVAL)
imputation_rule(df, x_stats, "max", "1/3")
imputation_rule(df, x_stats, "geom_mean", "1/3")
imputation_rule(df, x_stats, "mean", "1/2")

labels_use_control  Update labels according to control specifications

Description

[Stable]

Given a list of statistic labels and a list of control parameters, updates labels with a relevant control specification. For example, if control has element conf_level set to 0.9, the default label for statistic mean_ci will be updated to "Mean 90% CI". Any labels that are supplied via labels_custom will not be updated regardless of control.

Usage

labels_use_control(labels_default, control, labels_custom = NULL)

Arguments

labels_default (named character)
a named vector of statistic labels to modify according to the control specifications. Labels that are explicitly defined in labels_custom will not be affected.

control (named list)
list of control parameters to apply to adjust default labels.

labels_custom (named character)
named vector of labels that are customized by the user and should not be affected by control.

Value

A named character vector of labels with control specifications applied to relevant labels.

Examples

c control <- list(conf_level = 0.80, quantiles = c(0.1, 0.83), test_mean = 0.57)
g get_labels_from_stats(c("mean_ci", "quantiles", "mean_pval")) %>%
labels_use_control(control = control)
logistic_regression_cols

Logistic regression multivariate column layout function

Description

[Stable]
Layout-creating function which creates a multivariate column layout summarizing logistic regression results. This function is a wrapper for rtables::split_cols_by_multivar().

Usage

logistic_regression_cols(lyt, conf_level = 0.95)

Arguments

lyt (PreDataTableLayouts) layout that analyses will be added to.
conf_level (proportion) confidence level of the interval.

Value

A layout object suitable for passing to further layouting functions. Adding this function to an rtable layout will split the table into columns corresponding to statistics df, estimate, std_error, odds_ratio, ci, and pvalue.

logistic_summary_by_flag

Logistic regression summary table

Description

[Stable]
Constructor for content functions to be used in summarize_logistic() to summarize logistic regression results. This function is a wrapper for rtables::summarize_row_groups().

Usage

logistic_summary_by_flag(
  flag_var,
  na_str = default_na_str(),
  .indent_mods = NULL
)


Arguments

flag_var (string)
variable name identifying which row should be used in this content function.

na_str (string)
string used to replace all NA or empty values in the output.

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

Value

A content function.

month2day

Conversion of months to days

Description

[Stable]
Conversion of months to days. This is an approximative calculation because it considers each month as having an average of 30.4375 days.

Usage

month2day(x)

Arguments

x (numeric(1))
time in months.

Value

A numeric vector with the time in days.

Examples

x <- c(13.25, 8.15, 1, 2.834)
month2day(x)
### Description

**[Stable]**

Compares bivariate responses between two groups in terms of odds ratios along with a confidence interval.

### Usage

```r
estimate_odds_ratio(
  lyt,
  vars,
  variables = list(arm = NULL, strata = NULL),
  conf_level = 0.95,
  groups_list = NULL,
  na_str = default_na_str(),
  nested = TRUE,
  ...
)
```

```r
s_odds_ratio(
  df,
  .var,
  .ref_group,
  .in_ref_col,
  .df_row,
  variables = list(arm = NULL, strata = NULL),
  conf_level = 0.95,
  groups_list = NULL
)
```

```r
a_odds_ratio(
  df,
  .var,
  .ref_group,
  .in_ref_col,
  .df_row,
  variables = list(arm = NULL, strata = NULL),
)```
Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

vars (character)
variable names for the primary analysis variable to be iterated over.

variables (named list of string)
list of additional analysis variables.

conf_level (proportion)
confidence level of the interval.

groups_list (named list of character)
specifies the new group levels via the names and the levels that belong to it in
the character vectors that are elements of the list.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout
structure if possible (TRUE, the default) or as a new top-level element (FALSE).
Ignored if it would nest a split. underneath analyses, which is not allowed.

... arguments passed to s_odds_ratio().

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

table_names (character)
this can be customized in the case that the same vars are analyzed multiple
times, to avoid warnings from rtables.

.stats (character)
statistics to select for the table. Run get_stats("estimate_odds_ratio") to
see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on
the "auto" setting.

.labels (named character)
labels for the statistics (without indent).

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodi-
ﬁed default behavior. Can be negative.

df (data.frame)
data set containing all analysis variables.

.var (string)
single variable name that is passed by rtables when requested by a statistics
function.
odds_ratio

.ref_group (data.frame or vector)
the data corresponding to the reference group.

.in_ref_col (flag)
TRUE when working with the reference level, FALSE otherwise.

.df_row (data.frame)
data frame across all of the columns for the given row split.

Details
This function uses either logistic regression for unstratified analyses, or conditional logistic regression for stratified analyses. The Wald confidence interval with the specified confidence level is calculated.

Value
- estimate_odds_ratio() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_odds_ratio() to the table layout.

- s_odds_ratio() returns a named list with the statistics or_ci (containing est, lcl, and ucl) and n_tot.

- a_odds_ratio() returns the corresponding list with formatted rtables::CellValue().

Functions
- estimate_odds_ratio(): Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for rtables::analyze().

- s_odds_ratio(): Statistics function which estimates the odds ratio between a treatment and a control. A variables list with arm and strata variable names must be passed if a stratified analysis is required.

- a_odds_ratio(): Formatted analysis function which is used as afun in estimate_odds_ratio().

Note
For stratified analyses, there is currently no implementation for conditional likelihood confidence intervals, therefore the likelihood confidence interval is not yet available as an option. Besides, when rsp contains only responders or non-responders, then the result values will be NA, because no odds ratio estimation is possible.

See Also
Relevant helper function h_odds_ratio().
Examples

```r
set.seed(12)
dta <- data.frame(
  rsp = sample(c(TRUE, FALSE), 100, TRUE),
  grp = factor(rep(c("A", "B"), each = 50), levels = c("A", "B")),
  strata = factor(sample(c("C", "D"), 100, TRUE))
)

l <- basic_table() %>%
  split_cols_by(var = "grp", ref_group = "B") %>%
  estimate_odds_ratio(vars = "rsp")

build_table(l, df = dta)

# Unstratified analysis.
s_odds_ratio(
  df = subset(dta, grp == "A"),
  .var = "rsp",
  .ref_group = subset(dta, grp == "B"),
  .in_ref_col = FALSE,
  .df_row = dta
)

# Stratified analysis.
s_odds_ratio(
  df = subset(dta, grp == "A"),
  .var = "rsp",
  .ref_group = subset(dta, grp == "B"),
  .in_ref_col = FALSE,
  .df_row = dta,
  variables = list(arm = "grp", strata = "strata")
)

a_odds_ratio(
  df = subset(dta, grp == "A"),
  .var = "rsp",
  .ref_group = subset(dta, grp == "B"),
  .in_ref_col = FALSE,
  .df_row = dta
)
```

<table>
<thead>
<tr>
<th>prop_diff</th>
<th>Proportion difference</th>
</tr>
</thead>
</table>

Description

[Stable]
prop_diff

Usage

```
estimate_proportion_diff(  
  lyt,  
  vars,  
  variables = list(strata = NULL),  
  conf_level = 0.95,  
  method = c("waldcc", "wald", "cmh", "ha", "newcombe", "newcombecc", "strat_newcombe",  
    "strat_newcombecc"),  
  weights_method = "cmh",  
  na_str = default_na_str(),  
  nested = TRUE,  
  ...,  
  var_labels = vars,  
  show_labels = "hidden",  
  table_names = vars,  
  .stats = NULL,  
  .formats = NULL,  
  .labels = NULL,  
  .indent_mods = NULL  
)

s_proportion_diff(  
  df,  
  .var,  
  .ref_group,  
  .in_ref_col,  
  variables = list(strata = NULL),  
  conf_level = 0.95,  
  method = c("waldcc", "wald", "cmh", "ha", "newcombe", "newcombecc", "strat_newcombe",  
    "strat_newcombecc"),  
  weights_method = "cmh"  
)

a_proportion_diff(  
  df,  
  .var,  
  .ref_group,  
  .in_ref_col,  
  variables = list(strata = NULL),  
  conf_level = 0.95,  
  method = c("waldcc", "wald", "cmh", "ha", "newcombe", "newcombecc", "strat_newcombe",  
    "strat_newcombecc"),  
  weights_method = "cmh"  
)```
Arguments

lyt (PreDataTableLayouts) layout that analyses will be added to.

vars (character) variable names for the primary analysis variable to be iterated over.

variables (named list of string) list of additional analysis variables.

conf_level (proportion) confidence level of the interval.

method (string) the method used for the confidence interval estimation.

weights_method (string) weights method. Can be either "cmh" or "heuristic" and directs the way weights are estimated.

na_str (string) string used to replace all NA or empty values in the output.

nested (flag) whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split underneath analyses, which is not allowed.

... additional arguments for the lower level functions.

var_labels (character) variable labels.

show_labels (string) label visibility: one of "default", "visible" and "hidden".

table_names (character) this can be customized in the case that the same vars are analyzed multiple times, to avoid warnings from rtables.

.stats (character) statistics to select for the table. Run get_stats("estimate_proportion_diff") to see available statistics for this function.

.formats (named character or list) formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.

.labels (named character) labels for the statistics (without indent).

.indent_mods (named integer) indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

df (data.frame) data set containing all analysis variables.

.var (string) single variable name that is passed by rtables when requested by a statistics function.
prop_diff

.dataframe or vector
the data corresponding to the reference group.

(data.frame or vector)

TRUE when working with the reference level, FALSE otherwise.

TRUE

FALSE

Value

- estimate_proportion_diff() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_proportion_diff() to the table layout.

Functions

- estimate_proportion_diff(): Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for rtables::analyze().

- s_proportion_diff(): Statistics function estimating the difference in terms of responder proportion.

- a_proportion_diff(): Formatted analysis function which is used as afun in estimate_proportion_diff().

Note

When performing an unstratified analysis, methods "cmh", "strat_newcombe", and "strat_newcombecc" are not permitted.

See Also

d_proportion_diff()

Examples

```r
## "Mid" case: 4/4 respond in group A, 1/2 respond in group B.
nex <- 100 # Number of example rows
dta <- data.frame(
  "rsp" = sample(c(TRUE, FALSE), nex, TRUE),
  "grp" = sample(c("A", "B"), nex, TRUE),
  "f1" = sample(c("a1", "a2"), nex, TRUE),
  "f2" = sample(c("x", "y", "z"), nex, TRUE),
  stringsAsFactors = TRUE
)

l <- basic_table() %>%
  split_cols_by(var = "grp", ref_group = "B") %>%
estimate_proportion_diff(
  vars = "rsp",
  conf_level = 0.90,
  method = "ha"
)```
prune_occurrences

Occurrence table pruning

Description

[Stable]
Family of constructor and condition functions to flexibly prune occurrence tables. The condition functions always return whether the row result is higher than the threshold. Since they are of class CombinationFunction() they can be logically combined with other condition functions.

Usage

keep_rows(row_condition)

keep_content_rows(content_row_condition)
has_count_in_cols(atleast, ...)

has_count_in_any_col(atleast, ...)

has_fraction_in_cols(atleast, ...)

has_fraction_in_any_col(atleast, ...)

has_fractions_difference(atleast, ...)

has_counts_difference(atleast, ...)

Arguments

row_condition  (CombinationFunction)
condition function which works on individual analysis rows and flags whether these should be kept in the pruned table.

content_row_condition
(CombinationFunction)
condition function which works on individual first content rows of leaf tables and flags whether these leaf tables should be kept in the pruned table.

atleast  (numeric(1))
threshold which should be met in order to keep the row.

... arguments for row or column access, see rtables_access: either col_names (character) including the names of the columns which should be used, or alternatively col_indices (integer) giving the indices directly instead.

Value

• keep_rows() returns a pruning function that can be used with rtables::prune_table() to prune an rtables table.

• keep_content_rows() returns a pruning function that checks the condition on the first content row of leaf tables in the table.

• has_count_in_cols() returns a condition function that sums the counts in the specified column.

• has_count_in_any_col() returns a condition function that compares the counts in the specified columns with the threshold.

• has_fraction_in_cols() returns a condition function that sums the counts in the specified column, and computes the fraction by dividing by the total column counts.

• has_fraction_in_any_col() returns a condition function that looks at the fractions in the specified columns and checks whether any of them fulfill the threshold.

• has_fractions_difference() returns a condition function that extracts the fractions of each specified column, and computes the difference of the minimum and maximum.
• `has_counts_difference()` returns a condition function that extracts the counts of each specified column, and computes the difference of the minimum and maximum.

### Functions

- `keep_rows()`: Constructor for creating pruning functions based on a row condition function. This removes all analysis rows (TableRow) that should be pruned, i.e., don’t fulfill the row condition. It removes the sub-tree if there are no children left.

- `keep_content_rows()`: Constructor for creating pruning functions based on a condition for the (first) content row in leaf tables. This removes all leaf tables where the first content row does not fulfill the condition. It does not check individual rows. It then proceeds recursively by removing the sub tree if there are no children left.

- `has_count_in_cols()`: Constructor for creating condition functions on total counts in the specified columns.

- `has_count_in_any_col()`: Constructor for creating condition functions on any of the counts in the specified columns satisfying a threshold.

- `has_fraction_in_cols()`: Constructor for creating condition functions on total fraction in the specified columns.

- `has_fraction_in_any_col()`: Constructor for creating condition functions on any fraction in the specified columns.

- `has_fractions_difference()`: Constructor for creating condition function that checks the difference between the fractions reported in each specified column.

- `has_counts_difference()`: Constructor for creating condition function that checks the difference between the counts reported in each specified column.

### Note

Since most table specifications are worded positively, we name our constructor and condition functions positively, too. However, note that the result of `keep_rows()` says what should be pruned, to conform with the `rtables::prune_table()` interface.

### Examples

```r
tab <- basic_table() %>%
  split_cols_by("ARM") %>%
  split_rows_by("RACE") %>%
  split_rows_by("STRATA1") %>%
  summarize_row_groups() %>%
  analyze_vars("COUNTRY", .stats = "count_fraction") %>%
  build_table(DM)
```

```r
# 'keep_rows'
is_non_empty <- !CombinationFunction(all_zero_or_na)
prune_table(tab, keep_rows(is_non_empty))
```
# `keep_content_rows`
more_than_twenty <- has_count_in_cols(atleast = 20L, col_names = names(tab))
prune_table(tab, keep_content_rows(more_than_twenty))

more_than_one <- has_count_in_cols(atleast = 1L, col_names = names(tab))
prune_table(tab, keep_rows(more_than_one))

# `has_count_in_any_col`
any_more_than_one <- has_count_in_any_col(atleast = 1L, col_names = names(tab))
prune_table(tab, keep_rows(any_more_than_one))

# `has_fraction_in_cols`
more_than_five_percent <- has_fraction_in_cols(atleast = 0.05, col_names = names(tab))
prune_table(tab, keep_rows(more_than_five_percent))

# `has_fraction_in_any_col`
any_atleast_five_percent <- has_fraction_in_any_col(atleast = 0.05, col_names = names(tab))
prune_table(tab, keep_rows(more_than_five_percent))

# `has_fractions_difference`
more_than_five_percent_diff <- has_fractions_difference(atleast = 0.05, col_names = names(tab))
prune_table(tab, keep_rows(more_than_five_percent_diff))

more_than_one_diff <- has_counts_difference(atleast = 1L, col_names = names(tab))
prune_table(tab, keep_rows(more_than_one_diff))

reapply_varlabels --- Reapply variable labels

**Description**

This is a helper function that is used in tests.

**Usage**

reapply_varlabels(x, varlabels, ...)

---
response_biomarkers_subgroups

Tabulate biomarker effects on binary response by subgroup

Description

[Stable]

Tabulate the estimated effects of multiple continuous biomarker variables on a binary response endpoint across population subgroups.

Usage

```r
tabulate_rsp_biomarkers(
  df,
  vars = c("n_tot", "n_rsp", "prop", "or", "ci", "pval"),
  na_str = default_na_str(),
  .indentMods = 0L
)
```

Arguments

- `df` (data.frame) containing all analysis variables, as returned by `extract_rsp_biomarkers()`.
- `vars` (character) the names of statistics to be reported among:
  - `n_tot`: Total number of patients per group.
  - `n_rsp`: Total number of responses per group.
  - `prop`: Total response proportion per group.
  - `or`: Odds ratio.
  - `ci`: Confidence interval of odds ratio.
  - `pval`: p-value of the effect. Note, the statistics `n_tot`, `or` and `ci` are required.
response_biomarkers_subgroups

na_str (string)
string used to replace all NA or empty values in the output.

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

Details
These functions create a layout starting from a data frame which contains the required statistics. The tables are then typically used as input for forest plots.

Value
An rtables table summarizing biomarker effects on binary response by subgroup.

Note
In contrast to tabulate_rsp_subgroups() this tabulation function does not start from an input layout lyt. This is because internally the table is created by combining multiple subtables.

See Also
h_tab_rsp_one_biomarker() which is used internally, extract_rsp_biomarkers().

Examples
library(dplyr)
library(forcats)

adrs <- tern_ex_adrs
adrs_labels <- formatters::var_labels(adrs)

adrs_f <- adrs %>%
  filter(PARAMCD == "BESRSP") %>%
  mutate(rsp = AVALC == "CR")
formatters::var_labels(adrs_f) <- c(adrs_labels, "Response")

df <- extract_rsp_biomarkers(
  variables = list(
    rsp = "rsp",
    biomarkers = c("BMRKR1", "AGE"),
    covariates = "SEX",
    subgroups = "BMRKR2"
  ),
  data = adrs_f
)

## Table with default columns.
tabulate_rsp_biomarkers(df)
## Table with a manually chosen set of columns: leave out "pval", reorder.

```r
tab <- tabulate_rsp_biomarkers(
  df = df,
  vars = c("n_rsp", "ci", "n_tot", "prop", "or")
)
```

## Finally produce the forest plot.

```r
g_forest(tab, xlim = c(0.7, 1.4))
```

---

### rtable2gg

**Convert rtable objects to ggplot objects**

#### Description

**[Experimental]**

Given a rtables::rtable() object, performs basic conversion to a ggplot2::ggplot() object built using functions from the ggplot2 package. Any table titles and/or footnotes are ignored.

#### Usage

```r
table2gg(tbl, fontsize = 12, colwidths = NULL, lbl_col_padding = 0)
```

#### Arguments

- **tbl** *(VTableTree)*
  - rtables table object.
- **fontsize** *(numeric(1))*
  - font size.
- **colwidths** *(numeric or NULL)*
  - a vector of column widths. Each element’s position in colwidths corresponds to the column of tbl in the same position. If NULL, column widths are calculated according to maximum number of characters per column.
- **lbl_col_padding** *(numeric)*
  - additional padding to use when calculating spacing between the first (label) column and the second column of tbl. If colwidths is specified, the width of the first column becomes colwidths[1] + lbl_col_padding. Defaults to 0.

#### Value

- A ggplot object.
Examples

dta <- data.frame(
  ARM = rep(LETTERS[1:3], rep(6, 3)),
  AVISIT = rep(paste0("V", 1:3), 6),
  AVAL = c(9:1, rep(NA, 9))
)

lyt <- basic_table() %>%
  split_cols_by(var = "ARM") %>%
  split_rows_by(var = "AVISIT") %>%
  analyze_vars(vars = "AVAL")

tbl <- build_table(lyt, df = dta)

rtable2gg(tbl)

rtable2gg(tbl, fontsize = 15, colwidths = c(2, 1, 1, 1))

---

sas_na

Convert strings to NA

Description

[Stable]

SAS imports missing data as empty strings or strings with whitespaces only. This helper function can be used to convert these values to NAs.

Usage

sas_na(x, empty = TRUE, whitespaces = TRUE)

Arguments

x (factor or character)
  values for which any missing values should be substituted.
empty (flag)
  if TRUE, empty strings get replaced by NA.
whitespaces (flag)
  if TRUE, strings made from only whitespaces get replaced with NA.

Value

x with "" and/or whitespace-only values substituted by NA, depending on the values of empty and whitespaces.
Examples

```r
sas_na(c("1", "", "", "", "", "b"))
sas_na(factor(c("", "", "b")))

is.na(sas_na(c("1", "", "", "", "", "b")))
```

Description

**Occurrence table sorting**

Functions to score occurrence table subtables and rows which can be used in the sorting of occurrence tables.

Usage

- `score_occurrences(table_row)`
- `score_occurrences_cols(...)`
- `score_occurrences_subtable(...)`
- `score_occurrences_cont_cols(...)`

Arguments

- `table_row`: (TableRow) an analysis row in a occurrence table.
- `...`: arguments for row or column access, see `rtables_access`: either `col_names` (character) including the names of the columns which should be used, or alternatively `col_indices` (integer) giving the indices directly instead.

Value

- `score_occurrences()` returns the sum of counts across all columns of a table row.
- `score_occurrences_cols()` returns a function that sums counts across all specified columns of a table row.
- `score_occurrences_subtable()` returns a function that sums counts in each subtable across all specified columns.
- `score_occurrences_cont_cols()` returns a function that sums counts in the first content row in specified columns.
Functions

- `score_occurrences()`: Scoring function which sums the counts across all columns. It will fail if anything else but counts are used.
- `score_occurrences_cols()`: Scoring functions can be produced by this constructor to only include specific columns in the scoring. See `h_row_counts()` for further information.
- `score_occurrences_subtable()`: Scoring functions produced by this constructor can be used on subtables: They sum up all specified column counts in the subtable. This is useful when there is no available content row summing up these counts.
- `score_occurrences_cont_cols()`: Produces a score function for sorting table by summing the first content row in specified columns. Note that this is extending `rtables::cont_n_onecol()` and `rtables::cont_n_allcols()`.

See Also

- `h_row_first_values()`
- `h_row_counts()`

Examples

```r
lyt <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  analyze_num_patients(
    vars = "USUBJID",
    .stats = c("unique"),
    .labels = c("Total number of patients with at least one event")
  ) %>%
  split_rows_by("AEBODSYS", child_labels = "visible", nested = FALSE) %>%
  summarize_num_patients(
    var = "USUBJID",
    .stats = c("unique", "nonunique"),
    .labels = c(
      "Total number of patients with at least one event",
      "Total number of events"
    )
  ) %>%
  count_occurrences(vars = "AEDECOD")

tbl <- build_table(lyt, tern_ex_aadae, alt_counts_df = tern_ex_adsl) %>%
  prune_table()

tbl_sorted <- tbl %>%
  sort_at_path(path = c("AEBODSYS", "x", "AEDECOD"), scorefun = score_occurrences)

tbl_sorted

score_cols_a_and_b <- score_occurrences_cols(col_names = c("A: Drug X", "B: Placebo"))

# Note that this here just sorts the AEDECOD inside the AEBODSYS. The AEBODSYS are not sorted.
# That would require a second pass of `sort_at_path`. 
```
split_cols_by_groups

Split columns by groups of levels

Description

[Stable]

Usage

split_cols_by_groups(lyt, var, groups_list = NULL, ref_group = NULL, ...)

Arguments

- **lyt** (PreDataTableLayouts)
  layout that analyses will be added to.
- **var** (string)
  single variable name that is passed by rtables when requested by a statistics function.
- **groups_list** (named list of character)
  specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.
- **ref_group** (data.frame or vector)
  the data corresponding to the reference group.
- **...**
  additional arguments to rtables::split_cols_by() in order. For instance, to control formats (format), add a joint column for all groups (incl_all).

Value

A layout object suitable for passing to further laying out functions. Adding this function to an rtable layout will add a column split including the given groups to the table layout.
See Also

rtables::split_cols_by()

Examples

# 1 - Basic use

# Without group combination `split_cols_by_groups` is
# equivalent to [rtables::split_cols_by()].

basic_table() %>%
  split_cols_by_groups("ARM") %>%
  add_colcounts() %>%
  analyze("AGE") %>%
  build_table(DM)

# Add a reference column.

basic_table() %>%
  split_cols_by_groups("ARM", ref_group = "B: Placebo") %>%
  add_colcounts() %>%
  analyze(
    "AGE",
    afun = function(x, .ref_group, .in_ref_col) {
      if (.in_ref_col) {
        in_rows("Diff Mean" = rcell(NULL))
      } else {
        in_rows("Diff Mean" = rcell(mean(x) - mean(.ref_group), format = "xx.xx"))
      }
    }
  ) %>%
  build_table(DM)

# 2 - Adding group specification

# Manual preparation of the groups.

groups <- list(
  "Arms A+B" = c("A: Drug X", "B: Placebo"),
  "Arms A+C" = c("A: Drug X", "C: Combination")
)

# Use of split_cols_by_groups without reference column.

basic_table() %>%
  split_cols_by_groups("ARM", groups) %>%
  add_colcounts() %>%
  analyze("AGE") %>%
  build_table(DM)

# Including differentiated output in the reference column.

basic_table() %>%
  split_cols_by_groups("ARM", groups_list = groups, ref_group = "Arms A+B") %>%
  analyze(
    "AGE",
    afun = function(x, .ref_group, .in_ref_col) {
      if (.in_ref_col) {
        in_rows("Diff Mean" = rcell(NULL))
      } else {
        in_rows("Diff Mean" = rcell(mean(x) - mean(.ref_group), format = "xx.xx"))
      }
    }
  ) %>%
  build_table(DM)
if (.in_ref_col) {
    in_rows("Diff. of Averages" = rcell(NULL))
} else {
    in_rows("Diff. of Averages" = rcell(mean(x) - mean(.ref_group), format = "xx.xx"))
}
)
) %>%
build_table(DM)

# 3 - Binary list dividing factor levels into reference and treatment

# `combine_groups` defines reference and treatment.
groups <- combine_groups(
  fct = DM$ARM,
  ref = c("A: Drug X", "B: Placebo")
)
groups

# Use group definition without reference column.
basic_table() %>%
  split_cols_by_groups("ARM", groups_list = groups) %>%
  add_colcounts() %>%
  analyze("AGE") %>%
  build_table(DM)

# Use group definition with reference column (first item of groups).
basic_table() %>%
  split_cols_by_groups("ARM", groups, ref_group = names(groups)[1]) %>%
  add_colcounts() %>%
  analyze("AGE",
    afun = function(x, .ref_group, .in_ref_col) {
      if (.in_ref_col) {
        in_rows("Diff Mean" = rcell(NULL))
      } else {
        in_rows("Diff Mean" = rcell(mean(x) - mean(.ref_group), format = "xx.xx"))
      }
    }
  ) %>%
  build_table(DM)

---

**Description**

[Deprecated]

Stack grobs as a new grob with 1 column and multiple rows layout.
Usage

stack_grobs(
    ..., 
    grobs = list(...), 
    padding = grid::unit(2, "line"), 
    vp = NULL, 
    gp = NULL, 
    name = NULL 
)

Arguments

... grobs.
grobs (list of grob)
a list of grobs.
padding (grid::unit)
unit of length 1, space between each grob.
vp (viewport or NULL)
a viewport() object (or NULL).
gp (gpar)
a gpar() object.
name (string)
a character identifier for the grob.

Value

A grob.

Examples

library(grid)

g1 <- circleGrob(gp = gpar(col = "blue"))
g2 <- circleGrob(gp = gpar(col = "red"))
g3 <- textGrob("TEST TEXT")
grid.newpage()
grid.draw(stack_grobs(g1, g2, g3))
showViewport()

grid.newpage()
pushViewport(viewport(layout = grid.layout(1, 2)))
vp1 <- viewport(layout.pos.row = 1, layout.pos.col = 2)
g1 <- circleGrob(gp = gpar(col = "red"))
g2 <- textGrob("TEST TEXT")
g3 <- textGrob("TEST TEXT")
grid.draw(stack_grobs(g1, g2, g3, vp = vp1, name = "test"))
showViewport()
grid.ls(grobs = TRUE, viewports = TRUE, print = FALSE)
Confidence interval for mean

Description

[Stable]
Convenient function for calculating the mean confidence interval. It calculates the arithmetic as well as the geometric mean. It can be used as a ggplot helper function for plotting.

Usage

```r
stat_mean_ci(
  x,
  conf_level = 0.95,
  na.rm = TRUE,
  n_min = 2,
  gg_helper = TRUE,
  geom_mean = FALSE
)
```

Arguments

- `x` (numeric) vector of numbers we want to analyze.
- `conf_level` (proportion) confidence level of the interval.
- `na.rm` (flag) whether NA values should be removed from `x` prior to analysis.
- `n_min` (numeric(1)) a minimum number of non-missing `x` to estimate the confidence interval for mean.
- `gg_helper` (flag) whether output should be aligned for use with ggplots.
- `geom_mean` (flag) whether the geometric mean should be calculated.

Value

A named vector of values `mean_ci_lwr` and `mean_ci_upr`.

Examples

```r
stat_mean_ci(sample(10), gg_helper = FALSE)

p <- ggplot2::ggplot(mtcars, ggplot2::aes(cyl, mpg)) +
  ggplot2::geom_point()
```
stat_mean_pval

p + ggplot2::stat_summary(
  fun.data = stat_mean_ci,
  geom = "errorbar"
)

p + ggplot2::stat_summary(
  fun.data = stat_mean_ci,
  fun.args = list(conf_level = 0.5),
  geom = "errorbar"
)

p + ggplot2::stat_summary(
  fun.data = stat_mean_ci,
  fun.args = list(conf_level = 0.5, geom_mean = TRUE),
  geom = "errorbar"
)

---

stat_mean_pval  p-Value of the mean

Description

[Stable]
Convenient function for calculating the two-sided p-value of the mean.

Usage

stat_mean_pval(x, na.rm = TRUE, n_min = 2, test_mean = 0)

Arguments

x  (numeric)
vector of numbers we want to analyze.

na.rm  (flag)
whether NA values should be removed from x prior to analysis.

n_min  (numeric(1))
a minimum number of non-missing x to estimate the p-value of the mean.

test_mean  (numeric(1))
mean value to test under the null hypothesis.

Value

A p-value.
Examples

stat_mean_pval(sample(10))

stat_mean_pval(rnorm(10), test_mean = 0.5)

Description

[Stable]
Convenient function for calculating the median confidence interval. It can be used as a ggplot helper function for plotting.

Usage

stat_median_ci(x, conf_level = 0.95, na.rm = TRUE, gg_helper = TRUE)

Arguments

x (numeric)
vector of numbers we want to analyze.

conf_level (proportion)
confidence level of the interval.

na.rm (flag)
whether NA values should be removed from x prior to analysis.

gg_helper (flag)
whether output should be aligned for use with ggplots.

Details

This function was adapted from DescTools/versions/0.99.35/source

Value

A named vector of values median_ci_lwr and median_ci_upr.

Examples

stat_median_ci(sample(10), gg_helper = FALSE)

p <- ggplot2::ggplot(mtcars, ggplot2::aes(cyl, mpg)) +
ggplot2::geom_point()

p + ggplot2::stat_summary(
  fun.data = stat_median_ci,
  geom = "errorbar"
)
Function for calculating the proportion (or risk) difference and confidence interval between arm X (reference group) and arm Y. Risk difference is calculated by subtracting cumulative incidence in arm Y from cumulative incidence in arm X.

Usage

```r
stat_propdiff_ci(
  x,
  y,
  N_x,
  N_y,
  list_names = NULL,
  conf_level = 0.95,
  pct = TRUE
)
```

Arguments

- **x** (list of integer): list of number of occurrences in arm X (reference group).
- **y** (list of integer): list of number of occurrences in arm Y. Must be of equal length to x.
- **N_x** (numeric(1)): total number of records in arm X.
- **N_y** (numeric(1)): total number of records in arm Y.
- **list_names** (character): names of each variable/level corresponding to pair of proportions in x and y. Must be of equal length to x and y.
- **conf_level** (proportion): confidence level of the interval.
- **pct** (flag): whether output should be returned as percentages. Defaults to TRUE.

Value

List of proportion differences and CIs corresponding to each pair of number of occurrences in x and y. Each list element consists of 3 statistics: proportion difference, CI lower bound, and CI upper bound.
strata_normal_quantile

**See Also**

Split function `add_riskdiff()` which, when used as `split_fun` within `rtables::split_cols_by()` with `riskdiff` argument is set to `TRUE` in subsequent analyze functions, adds a column containing proportion (risk) difference to an `rtables` layout.

**Examples**

```r
stat_propdiff_ci(
  x = list(0.375), y = list(0.01), N_x = 5, N_y = 5, list_names = "x", conf_level = 0.9
)

df = rnorm(100)
df[1:50] <- 0

stat_propdiff_ci( 
  x = df, y = df, N_x = 10, N_y = 10, 
  list_names = "vanilla", conf_level = 0.9, 
  split_fun = add_riskdiff(), 
  split_base = 2
)
```

---

**strata_normal_quantile**

*Helper function for the estimation of stratified quantiles*

**Description**

[Stable]

This function wraps the estimation of stratified percentiles when we assume the approximation for large numbers. This is necessary only in the case proportions for each strata are unequal.

**Usage**

`strata_normal_quantile(vars, weights, conf_level)`

**Arguments**

- `vars` (character) variable names for the primary analysis variable to be iterated over.
- `weights` (numeric or `NULL`) weights for each level of the strata. If `NULL`, they are estimated using the iterative algorithm proposed in Yan and Su (2010) that minimizes the weighted squared length of the confidence interval.
- `conf_level` (proportion) confidence level of the interval.

**Value**

Stratified quantile.

**See Also**

`prop_strat_wilson()`
**Examples**

```r
strata_data <- table(data.frame(
    "f1" = sample(c(TRUE, FALSE), 100, TRUE),
    "f2" = sample(c("x", "y", "z"), 100, TRUE),
stringsAsFactors = TRUE
))
ns <- colSums(strata_data)
est <- strata_data["TRUE", ] / ns
vars <- ests * (1 - ests) / ns
weights <- rep(1 / length(ns), length(ns))
strata_normal_quantile(vars, weights, 0.95)
```

**summarize_colvars**  
*Summarize variables in columns*

**Description**

[Stable]

This analyze function uses the S3 generic function `s_summary()` to summarize different variables that are arranged in columns. Additional standard formatting arguments are available. It is a minimal wrapper for `rtables::analyze_colvars()`. The latter function is meant to add different analysis methods for each column variables as different rows. To have the analysis methods as column labels, please refer to `analyze_vars_in_cols()`.

**Usage**

```r
summarize_colvars(
    lyt,
    ...,  
    na_str = default_na_str(),
    .stats = c("n", "mean_sd", "median", "range", "count_fraction"),
    .formats = NULL,
    .labels = NULL,
    .indent_mods = NULL
)
```

**Arguments**

- `lyt` *(PreDataTableLayouts)*
  layout that analyses will be added to.
- `...` arguments passed to `s_summary()`.
- `na_str` *(string)*
  string used to replace all NA or empty values in the output.
- `stats` *(character)*
  statistics to select for the table.
summarize_colvars

.formats (named character or list)
  formats for the statistics. See Details in analyze_vars for more information on
  the "auto" setting.

.labels (named character)
  labels for the statistics (without indent).

.indent_mods (named vector of integer)
  indent modifiers for the labels. Each element of the vector should be a name-
  value pair with name corresponding to a statistic specified in .stats and value
  the indentation for that statistic’s row label.

Value

A layout object suitable for passing to further layouting functions, or to rtables::build_table().
Adding this function to an rtable layout will summarize the given variables, arrange the output in
columns, and add it to the table layout.

See Also

rtables::split_cols_by_multivar() and analyze_colvars_functions.

Examples

dta_test <- data.frame(
  USUBJID = rep(1:6, each = 3),
  PARAMCD = rep("lab", 6 * 3),
  AVISIT = rep(paste0("V", 1:3), 6),
  ARM = rep(LETTERS[1:3], rep(6, 3)),
  AVAL = c(9:1, rep(NA, 9)),
  CHG = c(1:9, rep(NA, 9))
)

## Default output within a `rtables` pipeline.
basic_table() %>%
  split_cols_by("ARM") %>%
  split_rows_by("AVISIT") %>%
  split_cols_by_multivar(vars = c("AVAL", "CHG")) %>%
  summarize_colvars() %>%
  build_table(dta_test)

## Selection of statistics, formats and labels also work.
basic_table() %>%
  split_cols_by("ARM") %>%
  split_rows_by("AVISIT") %>%
  split_cols_by_multivar(vars = c("AVAL", "CHG")) %>%
  summarize_colvars(
    .stats = c("n", "mean_sd"),
    .formats = c("mean_sd" = "xx.x, xx.x"),
    .labels = c(n = "n", mean_sd = "Mean, SD")
  ) %>%
  build_table(dta_test)
## Use arguments interpreted by `s_summary`.

```r
basic_table() %>%
  split_cols_by("ARM") %>%
  split_rows_by("AVISIT") %>%
  split_cols_by_multivar(vars = c("AVAL", "CHG")) %>%
  summarize_colvars(na.rm = FALSE) %>%
  build_table(dta_test)
```

---

### summarize_functions

**Summarize functions**

#### Description

These functions are wrappers for `rtables::summarize_row_groups()`, applying corresponding tern content functions to add summary rows to a given table layout:

#### Details

- `add_rowcounts()`
- `estimate_multinomial_response()` (with `rtables::analyze()`)
- `h_tab_one_biomarker()` (probably to deprecate)
- `logistic_summary_by_flag()`
- `summarize_num_patients()`
- `summarize_occurrences()`
- `summarize_occurrences_by_grade()`
- `summarize_patients_events_in_cols()`
- `summarize_patients_exposure_in_cols()`
- `tabulate_rsp_subgroups()`

Additionally, the `summarize_coxreg()` function utilizes `rtables::summarize_row_groups()` (in combination with several other `rtables` functions like `rtables::analyze_colvars()`) to output a Cox regression summary table.

#### See Also

- `analyze_functions` for functions which are wrappers for `rtables::analyze()`.
- `analyze_colvars_functions` for functions that are wrappers for `rtables::analyze_colvars()`.
**summarize_logistic**  
*Multivariate logistic regression table*

**Description**

[Stable]

Layout-creating function which summarizes a logistic variable regression for binary outcome with categorical/continuous covariates in model statement. For each covariate category (if categorical) or specified values (if continuous), present degrees of freedom, regression parameter estimate and standard error (SE) relative to reference group or category. Report odds ratios for each covariate category or specified values and corresponding Wald confidence intervals as default but allow user to specify other confidence levels. Report p-value for Wald chi-square test of the null hypothesis that covariate has no effect on response in model containing all specified covariates. Allow option to include one two-way interaction and present similar output for each interaction degree of freedom.

**Usage**

```r
summarize_logistic(
  lyt,
  conf_level,
  drop_and_remove_str = "",
  .indent.mods = NULL
)
```

**Arguments**

- `lyt` (PreDataTableLayouts) layout that analyses will be added to.
- `conf_level` (proportion) confidence level of the interval.
- `drop_and_remove_str` (string) string to be dropped and removed.
- `.indent.mods` (named integer) indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

**Value**

A layout object suitable for passing to further layouting functions, or to `rtables::build_table()`. Adding this function to an rtable layout will add a logistic regression variable summary to the table layout.

**Note**

For the formula, the variable names need to be standard data.frame column names without special characters.
Examples

```r
library(dplyr)
library(broom)

adrs_f <- tern_ex_adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(RACE %in% c("ASIAN", "WHITE", "BLACK OR AFRICAN AMERICAN")) %>%
  mutate(
    Response = case_when(AVALC %in% c("PR", "CR") ~ 1, TRUE ~ 0),
    RACE = factor(RACE),
    SEX = factor(SEX)
  )

formatters::var_labels(adrs_f) <- c(formatters::var_labels(tern_ex_adrs), Response = "Response")

mod1 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE")
  )
)

mod2 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE"),
    interaction = "AGE"
  )
)

df <- tidy(mod1, conf_level = 0.99)
df2 <- tidy(mod2, conf_level = 0.99)

# flagging empty strings with "_"

df <- df_explicit_na(df, na_level = "_")
df2 <- df_explicit_na(df2, na_level = "_")

result1 <- basic_table() %>%
  summarize_logistic(
    conf_level = 0.95,
    drop_and_remove_str = "_"
  ) %>%
  build_table(df = df)
result1

result2 <- basic_table() %>%
  summarize_logistic(
    conf_level = 0.95,
    drop_and_remove_str = "_"
  ) %>%
  build_table(df = df2)
```
summarize_num_patients

Number of patients

Description

[Stable]
Count the number of unique and non-unique patients in a column (variable).

Usage

analyze_num_patients(
  lyt,
  vars,
  required = NULL,
  count_by = NULL,
  unique_count_suffix = TRUE,
  na_str = default_na_str(),
  nested = TRUE,
  .stats = NULL,
  .formats = NULL,
  .labels = c(unique = "Number of patients with at least one event", nonunique = "Number of events"),
  show_labels = c("default", "visible", "hidden"),
  .indent_mods = 0L,
  riskdiff = FALSE,
  ...
)

summarize_num_patients(
  lyt,
  var,
  required = NULL,
  count_by = NULL,
  unique_count_suffix = TRUE,
  na_str = default_na_str(),
  .stats = NULL,
  .formats = NULL,
  .labels = c(unique = "Number of patients with at least one event", nonunique = "Number of events"),
  .indent_mods = 0L,
  riskdiff = FALSE,
  ...
)
summarize_num_patients

s_num_patients(
x,
labelstr,
.N_col,
count_by = NULL,
unique_count_suffix = TRUE)

s_num_patients_content(
df,
labelstr = "", .N_col, .var,
required = NULL,
count_by = NULL,
unique_count_suffix = TRUE)

Arguments

lyt (PreDataTableLayouts)
layout that analyses will be added to.

vars (character)
variable names for the primary analysis variable to be iterated over.

required (character or NULL)
optional, name of a variable that is required to be non-missing.

count_by (vector)
optional vector of any type to be combined with x when counting nonunique records.

unique_count_suffix (flag)
whether the "(n)" suffix should be added to unique_count labels. Defaults to TRUE.

na_str (string)
string used to replace all NA or empty values in the output.

nested (flag)
whether this layout instruction should be applied within the existing layout structure _if possible (TRUE, the default) or as a new top-level element (FALSE). Ignored if it would nest a split. underneath analyses, which is not allowed.

.stats (character)
statistics to select for the table. Run get_stats("summarize_num_patients") to see available statistics for this function.

.formats (named character or list)
formats for the statistics. See Details in analyze_vars for more information on the "auto" setting.
.labels (named character)
labels for the statistics (without indent).

show_labels (string)
label visibility: one of "default", "visible" and "hidden".

.indent_mods (named integer)
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

riskdiff (flag)
whether a risk difference column is present. When set to TRUE, add_riskdiff() must be used as split_fun in the prior column split of the table layout, specifying which columns should be compared. See stat_propdiff_ci() for details on risk difference calculation.

... additional arguments for the lower level functions.

x (character or factor)
vector of patient IDs.

labelstr (string)
label of the level of the parent split currently being summarized (must be present as second argument in Content Row Functions). See rtables::summarize_row_groups() for more information.

.N_col (integer(1))
column-wise N (column count) for the full column being analyzed that is typically passed by rtables.

df (data.frame)
data set containing all analysis variables.

.var, var (string)
single variable name that is passed by rtables when requested by a statistics function.

Details
In general, functions that start with analyze* are expected to work like rtables::analyze(), while functions that start with summarize* are based upon rtables::summarize_row_groups(). The latter provides a value for each dividing split in the row and column space, but, being it bound to the fundamental splits, it is repeated by design in every page when pagination is involved.

Value

- analyze_num_patients() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_num_patients_content() to the table layout.

- summarize_num_patients() returns a layout object suitable for passing to further layouting functions, or to rtables::build_table(). Adding this function to an rtable layout will add formatted rows containing the statistics from s_num_patients_content() to the table layout.
summarize_num_patients

- `s_num_patients()` returns a named list of 3 statistics:
  - unique: Vector of counts and percentages.
  - nonunique: Vector of counts.
  - unique_count: Counts.

- `s_num_patients_content()` returns the same values as `s_num_patients()`.

Functions

- `analyze_num_patients()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::analyze()`.
- `summarize_num_patients()`: Layout-creating function which can take statistics function arguments and additional format arguments. This function is a wrapper for `rtables::summarize_row_groups()`.
- `s_num_patients()`: Statistics function which counts the number of unique patients, the corresponding percentage taken with respect to the total number of patients, and the number of non-unique patients.
- `s_num_patients_content()`: Statistics function which counts the number of unique patients in a column (variable), the corresponding percentage taken with respect to the total number of patients, and the number of non-unique patients in the column.

Note

As opposed to `summarize_num_patients()`, this function does not repeat the produced rows.

Examples

```r
df <- data.frame(
  USUBJID = as.character(c(1, 2, 1, 4, NA, 6, 6, 8, 9)),
  AGE = c(10, 15, 10, 17, 8, 11, 11, 19, 17)
)

tbl <- basic_table() %>%
  split_cols_by("ARM") %>%
  add_colcounts() %>%
  analyze_num_patients("USUBJID", .stats = c("unique")) %>%
  build_table(df)

tbl

# Use the statistics function to count number of unique and nonunique patients.
s_num_patients(x = as.character(c(1, 1, 1, 2, 4, NA)), labelstr = "", .N_col = 6L)
s_num_patients(
  x = as.character(c(1, 1, 1, 2, 4, NA)),
  labelstr = "",
  .N_col = 6L,
  count_by = c(1, 1, 2, 1, 1)
)

# Count number of unique and non-unique patients.
```
```
df <- data.frame(
  USUBJID = as.character(c(1, 2, 1, 4, NA)),
  EVENT = as.character(c(10, 15, 10, 17, 8))
)
s_num_patients_content(df, .N_col = 5, .var = "USUBJID")

df_by_event <- data.frame(
  USUBJID = as.character(c(1, 2, 1, 4, NA)),
  EVENT = c(10, 15, 10, 17, 8)
)
s_num_patients_content(df_by_event, .N_col = 5, .var = "USUBJID", count_by = "EVENT")
```

---

**survival_biomarkers_subgroups**

*Tabulate biomarker effects on survival by subgroup*

**Description**

**[Stable]**

Tabulate the estimated effects of multiple continuous biomarker variables across population subgroups.

**Usage**

```r
tabulate_survival_biomarkers(
  df,
  vars = c("n_tot", "n_tot_events", "median", "hr", "ci", "pval"),
  groups_lists = list(),
  control = control_coxreg(),
  label_all = lifecycle::deprecated(),
  time_unit = NULL,
  na_str = default_na_str(),
  indent_mods = 0L
)
```

**Arguments**

- `df` *(data.frame)*
  containing all analysis variables, as returned by `extract_survival_biomarkers()`.

- `vars` *(character)*
  the names of statistics to be reported among:
  - `n_tot_events`: Total number of events per group.
  - `n_tot`: Total number of observations per group.
  - `median`: Median survival time.
  - `hr`: Hazard ratio.
• **ci**: Confidence interval of hazard ratio.
• **pval**: p-value of the effect. Note, one of the statistics n_tot and n_tot_events, as well as both hr and ci are required.

**groups_lists** *(named list of list)*
optionally contains for each `subgroups` variable a list, which specifies the new group levels via the names and the levels that belong to it in the character vectors that are elements of the list.

**control** *(list)*
a list of parameters as returned by the helper function `control_coxreg()`.

**label_all** *[Deprecated]*
please assign the `label_all` parameter within the `extract_survival_biomarkers()` function when creating `df`.

**time_unit** *(string)*
label with unit of median survival time. Default NULL skips displaying unit.

**na_str** *(string)*
string used to replace all NA or empty values in the output.

**.indent_mods** *(named integer)*
indent modifiers for the labels. Defaults to 0, which corresponds to the unmodified default behavior. Can be negative.

**Details**
These functions create a layout starting from a data frame which contains the required statistics. The tables are then typically used as input for forest plots.

**Value**
An `rtables` table summarizing biomarker effects on survival by subgroup.

**Functions**
- `tabulate_survival_biomarkers()`: Table-creating function which creates a table summarizing biomarker effects on survival by subgroup.

**Note**
In contrast to `tabulate_survival_subgroups()` this tabulation function does not start from an input layout `lyt`. This is because internally the table is created by combining multiple subtables.

**See Also**
- `h_tab_surv_one_biomarker()` which is used internally, `extract_survival_biomarkers()`.
Examples

```r
library(dplyr)

adtte <- tern_ex_adtte

# Save variable labels before data processing steps.
adtte_labels <- formatters::var_labels(adtte)

adtte_f <- adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(
    AVALU = as.character(AVALU),
    is_event = CNSR == 0
  )
labels <- c("AVALU" = adtte_labels[['AVALU']], "is_event" = "Event Flag")
formatters::var_labels(adtte_f)[names(labels)] <- labels

# Typical analysis of two continuous biomarkers `BMRKR1` and `AGE`,
# in multiple regression models containing one covariate `RACE`,
# as well as one stratification variable `STRATA1`. The subgroups
# are defined by the levels of `BMRKR2`.

df <- extractSurvivalBiomarkers(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    biomarkers = c("BMRKR1", "AGE"),
    strata = "STRATA1",
    covariates = "SEX",
    subgroups = "BMRKR2"
  ),
  label_all = "Total Patients",
  data = adtte_f
)
df

# Here we group the levels of `BMRKR2` manually.
df_grouped <- extractSurvivalBiomarkers(
  variables = list(
    tte = "AVAL",
    is_event = "is_event",
    biomarkers = c("BMRKR1", "AGE"),
    strata = "STRATA1",
    covariates = "SEX",
    subgroups = "BMRKR2"
  ),
  data = adtte_f,
  groups_lists = list(
    BMRKR2 = list(
      "low" = "LOW",
      "low/medium" = c("LOW", "MEDIUM"),
      "low/medium/high" = c("LOW", "MEDIUM", "HIGH")
    )
  )
)
df_grouped
## Table with default columns.
tabulate_survival_biomarkers(df)

## Table with a manually chosen set of columns: leave out "pval", reorder.
tab <- tabulate_survival_biomarkers(
  df = df,
  vars = c("n_tot_events", "ci", "n_tot", "median", "hr"),
  time_unit = as.character(adtte_f$AVALU[1])
)

## Finally produce the forest plot.
g_forest(tab, xlim = c(0.8, 1.2))

tidy.glm

---

**Custom tidy method for binomial GLM results**

**Description**

[Stable]

Helper method (for `broom::tidy()`) to prepare a data frame from a `glm` object with binomial family.

**Usage**

```r
## S3 method for class 'glm'
tidy(x, conf_level = 0.95, at = NULL, ...)
```

**Arguments**

- `x` *(glm)*
  - logistic regression model fitted by `stats::glm()` with "binomial" family.
- `conf_level` *(proportion)*
  - confidence level of the interval.
- `at` *(numeric or NULL)*
  - optional values for the interaction variable. Otherwise the median is used.
- `...`
  - additional arguments for the lower level functions.

**Value**

A data.frame containing the tidied model.
See Also

h_logistic_regression for relevant helper functions.

Examples

library(dplyr)
library(broom)

adrs_f <- tern_ex_adrs %>%
  filter(PARAMCD == "BESRSPI") %>%
  filter(RACE %in% c("ASIAN", "WHITE", "BLACK OR AFRICAN AMERICAN")) %>%
  mutate(
    Response = case_when(AVALC %in% c("PR", "CR") ~ 1, TRUE ~ 0),
    RACE = factor(RACE),
    SEX = factor(SEX)
  )
formatters::var_labels(adrs_f) <- c(formatters::var_labels(tern_ex_adrs), Response = "Response")
mod1 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE")
  )
)
mod2 <- fit_logistic(
  data = adrs_f,
  variables = list(
    response = "Response",
    arm = "ARMCD",
    covariates = c("AGE", "RACE"),
    interaction = "AGE"
  )
)

df <- tidy(mod1, conf_level = 0.99)
df2 <- tidy(mod2, conf_level = 0.99)

---

**tidy.step**

*Custom tidy method for STEP results*

**Description**

**[Stable]**

Tidy the STEP results into a tibble format ready for plotting.
tidy_coaxreg

Usage

```r
## S3 method for class 'step'
tidy(x, ...)
```

Arguments

- `x` (matrix)
  results from `fit_survival_step()`.
- `...` not used.

Value

A tibble with one row per STEP subgroup. The estimates and CIs are on the HR or OR scale, respectively. Additional attributes carry metadata also used for plotting.

See Also

- `g_step()` which consumes the result from this function.

Examples

```r
library(survival)
lung$sex <- factor(lung$sex)
vars <- list(
  time = "time",
  event = "status",
  arm = "sex",
  biomarker = "age"
)
step_matrix <- fit_survival_step(
  variables = vars,
  data = lung,
  control = c(control_coxph(), control_step(num_points = 10, degree = 2))
)
broom::tidy(step_matrix)
```
Usage

## S3 method for class 'summary.coxph'
 tidy(x, ...)

## S3 method for class 'coxreg.univar'
 tidy(x, ...)

## S3 method for class 'coxreg.multivar'
 tidy(x, ...)

Arguments

x (list)
result of the Cox regression model fitted by `fit_coxreg_univar()` (for univariate models) or `fit_coxreg_multivar()` (for multivariate models).

... additional arguments for the lower level functions.

Value

`broom::tidy()` returns:

- For `summary.coxph` objects, a data.frame with columns: `Pr(>|z|)`, `exp(coef)`, `exp(-coef)`, `lower .95`, `upper .95`, `level`, and `n`.
- For `coxreg.univar` objects, a data.frame with columns: `effect`, `term`, `term_label`, `level`, `n`, `hr`, `lcl`, `ucl`, `pval`, and `ci`.
- For `coxreg.multivar` objects, a data.frame with columns: `term`, `pval`, `term_label`, `hr`, `lcl`, `ucl`, `level`, and `ci`.

Functions

- `tidy(summary.coxph)`: Custom tidy method for `survival::coxph()` summary results. Tidy the `survival::coxph()` results into a data.frame to extract model results.
- `tidy(coxreg.univar)`: Custom tidy method for a univariate Cox regression. Tidy up the result of a Cox regression model fitted by `fit_coxreg_univar()`.
- `tidy(coxreg.multivar)`: Custom tidy method for a multivariate Cox regression. Tidy up the result of a Cox regression model fitted by `fit_coxreg_multivar()`.

See Also

cox_regression

Examples

library(survival)
library(broom)
set.seed(1, kind = "Mersenne-Twister")
```r
dta_bladder <- with(
  data = bladder[bladder$enum < 5, ],
  data.frame(
    time = stop,
    status = event,
    armcd = as.factor(rx),
    covar1 = as.factor(enum),
    covar2 = factor(
      sample(as.factor(enum)),
      levels = 1:4, labels = c("F", "F", "M", "M")
    )
  )
)

labels <- c("armcd" = "ARM", "covar1" = "A Covariate Label", "covar2" = "Sex (F/M)"
)
formatters::var_labels(dta_bladder)[names(labels)] <- labels

da_bladder$age <- sample(20:60, size = nrow(dta_bladder), replace = TRUE)

formula <- "survival::Surv(time, status) ~ armcd + covar1"
msum <- summary(coxph(stats::as.formula(formula), data = dta_bladder))
tidy(msum)

## Cox regression: arm + 1 covariate.
mod1 <- fit_coxreg_univar(
  variables = list(
    time = "time", event = "status", arm = "armcd",
    covariates = "covar1"
  ),
  data = dta_bladder,
  control = control_coxreg(conf_level = 0.91)
)

## Cox regression: arm + 1 covariate + interaction, 2 candidate covariates.
mod2 <- fit_coxreg_univar(
  variables = list(
    time = "time", event = "status", arm = "armcd",
    covariates = c("covar1", "covar2"
  ),
  data = dta_bladder,
  control = control_coxreg(conf_level = 0.91, interaction = TRUE)
)

tidy(mod1)
tidy(mod2)

multivar_model <- fit_coxreg_multivar(
  variables = list(
    time = "time", event = "status", arm = "armcd",
    covariates = c("covar1", "covar2"
  ),
  data = dta_bladder
)
broom::tidy(multivar_model)
```

### to_n

**Replicate entries of a vector if required**

**Description**

[Stable]

Replicate entries of a vector if required.

**Usage**

```
to_n(x, n)
```

**Arguments**

- **x** *(numeric)* vector of numbers we want to analyze.
- **n** *(integer(1))* number of entries that are needed.

**Value**

x if it has the required length already or is NULL, otherwise if it is scalar the replicated version of it with n entries.

**Note**

This function will fail if x is not of length n and/or is not a scalar.

### to_string_matrix

**Convert table into matrix of strings**

**Description**

[Stable]

Helper function to use mostly within tests. with_spaces parameter allows to test not only for content but also indentation and table structure. print_txt_to_copy instead facilitate the testing development by returning a well formatted text that needs only to be copied and pasted in the expected output.
to_string_matrix

Usage

to_string_matrix(
  x,
  widths = NULL,
  max_width = NULL,
  hsep = formatters::default_hsep(),
  with_spaces = TRUE,
  print_txt_to_copy = FALSE
)

Arguments

x (VTableTree)
  rtables table object.

widths (numeric or NULL)
  Proposed widths for the columns of x. The expected length of this numeric
  vector can be retrieved with ncol(x) + 1 as the column of row names must also
  be considered.

max_width (integer(1), string or NULL)
  width that title and footer (including footnotes) materials should be word-wrapped
  to. If NULL, it is set to the current print width of the session (getOption("width")).
  If set to "auto", the width of the table (plus any table inset) is used. Parameter
  is ignored if tf_wrap = FALSE.

hsep (string)
  character to repeat to create header/body separator line. If NULL, the object value
  will be used. If " ", an empty separator will be printed. See default_hsep() for
  more information.

with_spaces (flag)
  whether the tested table should keep the indentation and other relevant spaces.

print_txt_to_copy (flag)
  utility to have a way to copy the input table directly into the expected variable
  instead of copying it too manually.

Value

A matrix of strings. If print_txt_to_copy = TRUE the well formatted printout of the table will
be printed to console, ready to be copied as a expected value.

Examples

tbl <- basic_table() %>%
  split_rows_by("SEX") %>%
  split_cols_by("ARM") %>%
  analyze("AGE") %>%
  build_table(tern_ex_adsl)

to_string_matrix(tbl, widths = ceiling(propose_column_widths(tbl) / 2))
univariate

Univariate formula special term

Description

[Stable]

The special term `univariate` indicate that the model should be fitted individually for every variable included in univariate.

Usage

`univariate(x)`

Arguments

- `x` (character)

  a vector of variable names separated by commas.

Details

If provided alongside with pairwise specification, the model `y ~ ARM + univariate(SEX, AGE, RACE)` lead to the study and comparison of the models

- `y ~ ARM`
- `y ~ ARM + SEX`
- `y ~ ARM + AGE`
- `y ~ ARM + RACE`

Value

When used within a model formula, produces univariate models for each variable provided.
update_weights_strat_wilson

*Helper function for the estimation of weights for prop_strat_wilson()*

**Description**

[Stable]
This function wraps the iteration procedure that allows you to estimate the weights for each proportional strata. This assumes to minimize the weighted squared length of the confidence interval.

**Usage**

```r
update_weights_strat_wilson(
  vars,
  strata_qnorm,
  initial_weights,
  n_per_strata,
  max_iterations = 50,
  conf_level = 0.95,
  tol = 0.001
)
```

**Arguments**

- `vars` (numeric)
  normalized proportions for each strata.

- `strata_qnorm` (numeric(1))
  initial estimation with identical weights of the quantiles.

- `initial_weights` (numeric)
  initial weights used to calculate `strata_qnorm`. This can be optimized in the future if we need to estimate better initial weights.

- `n_per_strata` (numeric)
  number of elements in each strata.

- `max_iterations` (integer(1))
  maximum number of iterations to be tried. Convergence is always checked.

- `conf_level` (proportion)
  confidence level of the interval.

- `tol` (numeric(1))
  tolerance threshold for convergence.

**Value**

A list of 3 elements: `n_it`, `weights`, and `diff_v`.
See Also

For references and details see prop_strat_wilson().

Examples

```r
vs <- c(0.011, 0.013, 0.012, 0.014, 0.017, 0.018)
sq <- 0.674
ws <- rep(1 / length(vs), length(vs))
ns <- c(22, 18, 17, 17, 14, 12)
update_weights_strat_wilson(vs, sq, ws, ns, 100, 0.95, 0.001)
```

---

**utils_split_funs**  
Custom split functions

---

### Description

[Stable]

Collection of useful functions that are expanding on the core list of functions provided by `rtables`. See `rtables::custom_split_funs` and `rtables::make_split_fun()` for more information on how to make a custom split function. All these functions work with `rtables::split_rows_by()` argument `split_fun` to modify the way the split happens. For other split functions, consider consulting `rtables::split_funcs`.

### Usage

```
ref_group_position(position = "first")
level_order(order)
```

### Arguments

- **position**  
  (string or integer)
  position to use for the reference group facet. Can be "first", "last", or a specific position.

- **order**  
  (character or numeric)
  vector of ordering indices for the split facets.

### Value

- `ref_group_position()` returns an utility function that puts the reference group as first, last or at a certain position and needs to be assigned to `split_fun`.

- `level_order()` returns an utility function that changes the original levels’ order, depending on input `order` and split levels.
Functions

- `ref_group_position()`: Split function to place reference group facet at a specific position during post-processing stage.
- `level_order()`: Split function to change level order based on an integer vector or a character vector that represent the split variable’s factor levels.

See Also

`rtables::make_split_fun()`

Examples

```r
library(dplyr)

dat <- data.frame(
  x = factor(letters[1:5], levels = letters[5:1]),
  y = 1:5
)

# With rtables layout functions
basic_table() %>%
  split_cols_by("x", ref_group = "c", split_fun = ref_group_position("last")) %>%
  analyze("y") %>%
  build_table(dat)

# With tern layout functions
adtte_f <- tern_ex_adtte %>%
  filter(PARAMCD == "OS") %>%
  mutate(
    AVAL = day2month(AVAL),
    is_event = CNSR == 0
  )

basic_table() %>%
  split_cols_by(var = "ARMCD", ref_group = "ARM B", split_fun = ref_group_position("first")) %>%
  add_colcounts() %>%
  surv_time(
    vars = "AVAL",
    var_labels = "Survival Time (Months)",
    is_event = "is_event",
  ) %>%
  build_table(df = adtte_f)

basic_table() %>%
  split_cols_by(var = "ARMCD", ref_group = "ARM B", split_fun = ref_group_position(2)) %>%
  add_colcounts() %>%
  surv_time(
    vars = "AVAL",
    var_labels = "Survival Time (Months)",
    is_event = "is_event",
  ) %>%
```
build_table(df = adtte_f)

# level_order -------
# Even if default would bring ref_group first, the original order puts it last
basic_table() %>%
  split_cols_by("Species", split_fun = level_order(c(1, 3, 2))) %>%
  analyze("Sepal.Length") %>%
  build_table(iris)

# character vector
new_order <- level_order(levels(iris$Species)[c(1, 3, 2)])
basic_table() %>%
  split_cols_by("Species", ref_group = "virginica", split_fun = new_order) %>%
  analyze("Sepal.Length") %>%
  build_table(iris)
Index

!,,CombinationFunction-method (combination_function), 30
  * datasets
default_stats_formats_labels, 88
ex_data, 115
  * formatting functions 
extreme_format, 113
format_auto, 129
format_count_fraction, 130
format_count_fraction_fixed_dp, 131
format_count_fraction_lt10, 131
format_extreme_values, 132
format_extreme_values_ci, 133
format_fraction, 134
format_fraction_fixed_dp, 135
format_fraction_threshold, 136
format_sigfig, 137
format_xx, 138
formatting_functions, 128
&,,CombinationFunction,CombinationFunction-method (combination_function), 30
a_count_occurrences (count_occurrences), 49
a_count_occurrences_by_grade (count_occurrences_by_grade), 54
a_count_patients_with_event (count_patients_with_event), 59
a_count_patients_with_flags (count_patients_with_flags), 63
a_count_values (count_values_funs), 67
a_coxreg (cox_regression), 70
a_length_proportion (estimate_multinomial_rsp), 101
a_odds_ratio (odds_ratio), 231
a_proportion (estimate_proportions), 103
a_proportion_diff (prop_diff), 234
a_summary (analyze_variables), 13
a_summary(), 38
abnormal_by_worst_grade, 163
abnormal_by_worst_grade_worsen, 164, 225
add_riskdiff, 8
add_riskdiff(), 51, 56, 60, 64, 256, 264
add_rowcounts, 9
add_rowcounts(), 259
aes_label, 10
analyze_colvars_functions, 11, 12, 258, 259
analyze_functions, 11, 11, 259
analyze_num_patients (summarize_num_patients), 262
analyze_num_patients(), 11
analyze_patients_exposure_in_cols(), 11
analyze_variables, 13
analyze_vars, 129
analyze_vars (analyze_variables), 13
analyze_vars(), 12, 13, 24, 40, 88–90, 129
analyze_vars_in_cols, 21
analyze_vars_in_cols(), 11, 90, 227, 257
append_varlabels, 25
arrange_grobs, 27
as.rtable, 28
bland_altman, 29
broom::tidy(), 72, 174, 269, 272
c combination_function, 30
CombinationFunction (combination_function), 30
CombinationFunction(), 238
CombinationFunction-class (combination_function), 30
combine_counts, 32
combine_groups, 33
combine_groups(), 32
combine_levels, 34

281
h_col_indices, 168
h_count_cumulative, 169
h_cox_regression, 74, 119, 170
h_coxph_df
  (h_survival_duration_subgroups), 216
h_coxph_subgroups_df
  (h_survival_duration_subgroups), 216
h_coxph_subgroups_df(), 112
h_coxreg_extract_interaction
  (cox_regression_inter), 76
h_coxreg_inter_effect
  (cox_regression_inter), 76
h_coxreg_inter_estimations
  (cox_regression_inter), 76
h_coxreg_mult_cont_df
  (h_survival_biomarkers_subgroups), 213
h_coxreg_mult_cont_df(), 112, 214
h_coxreg_multivar_extract
  (h_cox_regression), 170
h_coxreg_multivar_formula
  (h_cox_regression), 170
h_coxreg_univar_extract
  (h_cox_regression), 170
h_coxreg_univar_extract()
  (h_cox_regression), 78
h_coxreg_univar_formulas
  (h_cox_regression), 170
h_data_plot, 173
h_decompose_gg, 175
h_format_row, 176
h_format_threshold(extreme_format), 113
h_g_ipp, 184
h_g_ipp()
  (extreme_format), 147
h_get_format_threshold
  (extreme_format), 113
h_get_interaction_vars
  (h_logistic_regression), 188
h_ggkm, 177
h_glm_inter_term_extract
  (h_logistic_regression), 188
h_glm_interaction_extract
  (h_logistic_regression), 188
h_glm_simple_term_extract
  (h_logistic_regression), 188
h_glm_simple_term_extract()
  (h_logistic_regression), 190
h_grob_coxph, 179
h_grob_median_surv, 180
h_grob_tbl_at_risk, 181
h_grob_y_annot, 183
h_interaction_coef_name
  (h_logistic_regression), 188
h_interaction_term_labels
  (h_logistic_regression), 188
h_km_layout, 186
h_logistic_inter_terms
  (h_logistic_regression), 188
h_logistic_mult_cont_df
  (h_response_biomarkers_subgroups), 201
h_logistic_mult_cont_df()
  (h_logistic_regression), 188, 270
h_logistic_simple_terms
  (h_logistic_regression), 188
h_map_for_count_abnormal, 192
h_odds_ratio, 194
h_odds_ratio()
  (h_response_subgroups), 233
h_odds_ratio_df
  (h_response_subgroups), 204
h_odds_ratio_subgroups_df
  (h_response_subgroups), 204
h_odds_ratio_subgroups_df()
  (h_logistic_regression), 110
h_or_cat_interaction
  (h_logistic_regression), 188
h_or_cat_interaction()
  (h_logistic_regression), 188
h_or_cont_interaction
  (h_logistic_regression), 188
h_or_cont_interaction()
  (h_logistic_regression), 190
h_or_interaction
  (h_logistic_regression), 188
h_or_interaction()
  (h_logistic_regression), 190
h_pkparam_sort, 195
h_prop_diff, 198
h_proportion_df
  (h_response_subgroups), 204
h_proportion_df()
  (h_logistic_regression), 188
h_proportion_subgroups_df
  (h_response_subgroups), 204
h_proportion_subgroups_df()
  (h_logistic_regression), 110
h_proportions, 106, 196
h_response_biomarkers_subgroups, 201
h_response_subgroups, 204
h_row_counts(), 247
h_row_first_values(), 247
h_rsp_to_logistic_variables
  (h_response_biomarkers_subgroups),
h_simple_term_labels
  (h_logistic_regression), 188
h_split_by_subgroups, 207
h_split_param, 209
h_stack_by_baskets, 210
h_step, 211
h_step_rsp_est (h_step), 211
h_step_rsp_formula (h_step), 211
h_step_survival_est (h_step), 211
h_step_trt_effect (h_step), 211
h_step_window (h_step), 211
h_surv_to_coxreg_variables
  (h_survival_biomarkers_subgroups), 213
h_survival_biomarkers_subgroups, 213
h_survival_duration_subgroups, 216
h_survtime_df
  (h_survival_duration_subgroups), 216
h_survtime_subgroups_df
  (h_survival_duration_subgroups), 216
h_survtime_subgroups_df(), 112
h_tab_one_biomarker, 220
h_tab_one_biomarker(), 259
h_tab_rsp_one_biomarker
  (h_response_biomarkers_subgroups), 201
h_tab_rsp_one_biomarker(), 221, 243
h_tab_surv_one_biomarker
  (h_survival_biomarkers_subgroups), 213
h_tab_surv_one_biomarker(), 221, 267
h_tbl_coxph_pairwise, 221
h_tbl_coxph_pairwise(), 179
h_tbl_medianSurv, 223
h_worsen_counter, 224
h_xticks, 225
has_count_in_any_col
  (prune_occurrences), 238
has_count_in_cols (prune_occurrences), 238
has_counts_difference
  (prune_occurrences), 238
has_fraction_in_any_col
  (prune_occurrences), 238
has_fractions_difference
  (prune_occurrences), 238
imputation_rule, 226
imputation_rule(), 23
incidence_rate, 44
individual_patient_plot (g_ipp), 145
kaplan_meier (g_km), 148
keep_content_rows (prune_occurrences), 238
keep_rows (prune_occurrences), 238
keep_rows(), 240
labeling::extended(), 150, 155, 178, 226
labels_use_control, 228
length(), 17
level_order (utils_split_funs), 278
levels(), 77
logistic_regression_cols, 229
logistic_summary_by_flag, 229
logistic_summary_by_flag(), 259
max(), 17
mean(), 17
median(), 17
min(), 17
month2day, 230
odds_ratio, 194, 231
or_clogit (h_odds_ratio), 194
or_clogit(), 194
or_glm (h_odds_ratio), 194
prop_agresti_coull (h_proportions), 196
prop_clopper_pearson (h_proportions), 196
prop_diff, 98, 234
prop_diff(), 199
prop_diff_cmh (h_prop_diff), 198
prop_diff_cmh(), 199
prop_diff_ha (h_prop_diff), 198
prop_diff_nc (h_prop_diff), 198
prop_diff_strat_nc (h_prop_diff), 198
prop_diff_wald (h_prop_diff), 198
prop_jeffreys (h_proportions), 196
prop_strat_wilson (h_proportions), 196
prop_strat_wilson(), 199, 256, 278
prop_wald (h_proportions), 196
prop_wilson (h_proportions), 196
prune_occurrences, 238

range_noinf(), 17
reapply_varlabels, 241
ref_group_position (utils_split_funs), 278
response_biomarkers_subgroups, 242
response_subgroups, 111
rtable2gg, 244
rtables::add_colevents (), 9
rtables::add_combo_levels(), 8, 140
rtables::analyse(), 11, 18, 37, 53, 57, 62, 66, 70, 73, 102, 106, 233, 237, 259, 264, 265
rtables::analyse_colvars(), 11, 12, 22, 24, 73, 257, 259
rtables::build_table(), 17, 18, 23, 37, 52, 57, 61, 65, 69, 73, 78, 102, 106, 233, 237, 258, 260, 264
rtables::CellValue(), 18, 52, 57, 61, 66, 69, 73, 102, 106, 233, 237
rtables::cont_n_allcols(), 247
rtables::cont_n_onecol(), 247
rtables::custom_split_funs, 278
rtables::make_split_fun(), 278, 279
rtables::prune_table(), 239, 240
rtables::rtable(), 143, 244
rtables::split_cols_by(), 8, 248, 249, 256
rtables::split_cols_by_multivar(), 229, 258
rtables::split_funcs, 278
rtables::split_rows_by(), 278
rtables::summarize_row_groups(), 9, 11, 12, 23, 24, 52, 53, 57, 73, 102, 221, 229, 259, 264, 265
rtables_access, 239, 246

s_bland_altman (bland_altman), 29
s_compare (compare_variables), 35
s_count_abnormal_by_baseline, 94
s_count_cumulative(), 95
s_count_missed_doses(), 96
s_count_occurrences
  (count_occurrences), 49

s_count_occurrences_by_grade
  (count_occurrences_by_grade), 54
s_count_occurrences_by_grade(), 167
s_count_patients_with_event
  (count_patients_with_event), 59
s_count_patients_with_flags
  (count_patients_with_flags), 63
s_count_values (count_values_funs), 67
s_coxph_pairwise(), 42
s_coxreg (cox_regression), 70
s_length_proportion
  (estimate_multinomial_rsp), 101
s_num_patients
  (summarize_num_patients), 262
s_num_patients_content
  (summarize_num_patients), 262
s_odds_ratio (odds_ratio), 231
s_proportion (estimate_proportions), 103
s_proportion(), 97, 102
s_proportion_diff (prop_diff), 234
s_proportion_diff(), 98
s_summary (analyze_variables), 13
s_summary(), 13, 37, 38, 40, 69, 227, 257
s_surv_time(), 48
s_surv_timepoint(), 49
s_test_proportion_diff(), 99, 111, 205
sas_na, 245
sas_na(), 92
score_occurrences, 246
score_occurrences_cols
  (score_occurrences), 246
score_occurrences_cont_cols
  (score_occurrences), 246
score_occurrences_subtable
  (score_occurrences), 246
set_default_na_str (default_na_str), 87
set_default_na_str(), 87
split_cols_by_groups, 248
sqrt(), 17
stack_grobs, 250
stat_mean_ci, 252
stat_mean_ci(), 17
stat_mean_pval, 253
stat_mean_pval(), 17
stat_median_ci, 254
stat_median_ci(), 17
stat_propdiff_ci, 255
INDEX

stat_propdiff_ci(), 8, 51, 56, 60, 65, 264
stats::as.formula(), 171
stats::binom.test(), 197
stats::glm(), 189, 194, 269
stats::IQR(), 17
stats::mantelhaen.test(), 199
stats::median(), 17
stats::prop.test(), 197, 199
stats::quantile(), 16, 17, 40, 80
stats::sd(), 17
strata_normal_quantile(), 256
strata_normal_quantile(), 197
sum(), 17
summarize_ancova(), 12
summarize_change(), 12
summarize_colvars, 257
summarize_colvars(), 12
summarize_coxreg(cox_regression), 70
summarize_coxreg(), 11, 259
summarize_functions, 11, 12, 259
summarize_logistic, 260
summarize_logistic(), 229
summarize_num_patients, 262
summarize_num_patients(), 259, 265
summarize_occurrences
  (count_occurrences), 49
summarize_occurrences(), 259
summarize_occurrences_by_grade
  (count_occurrences_by_grade), 54
summarize_occurrences_by_grade(), 259
summarize_patients_events_in_cols(), 259
summarize_patients_events_in_cols(), 11, 259
summary_formats
  (default_stats_formats_labels), 88
summary_labels
  (default_stats_formats_labels), 88
surv.time(), 12
surv.timepoint(), 12
survival::clogit(), 189, 194
survival::coxph(), 42, 43, 77, 113, 119, 150, 151, 171, 218, 222, 272
survival::summary.survfit(), 182
survival::survdiff(), 113, 218
survival::survfit(), 48, 49, 149, 174, 178, 181, 223
survival_biomarkers_subgroups, 266
survival_duration_subgroups, 113
tabulate_rsp_biomarkers
  (response_biomarkers_subgroups), 242
tabulate_rsp_subgroups(), 11, 98, 243, 259
tabulate_survival_biomarkers
  (survival_biomarkers_subgroups), 266
tabulate_survival_biomarkers(), 112
tabulate_survival_biomarkers(), 11, 99, 267
tern(tern-package), 7
tern-package, 7
tern_default_formats
  (default_stats_formats_labels), 88
tern_default_labels
  (default_stats_formats_labels), 88
tern_default_stats
  (default_stats_formats_labels), 88
tern_ex_aade(ex_data), 115
tern_ex_adlb(ex_data), 115
tern_ex_adpp(ex_data), 115
tern_ex_adrs(ex_data), 115
tern_ex_adsd(ex_data), 115
tern_ex_adtte(ex_data), 115
test_proportion_diff(), 12
tidy.coxreg.multivar(tidy_coxreg), 271
tidy.coxreg.univar(tidy_coxreg), 271
tidy.glm, 269
tidy.step, 270
tidy.step(), 158
tidy.summary.coxph(tidy_coxreg), 271
tidy_coxreg, 74, 271
to_n, 274
to_string_matrix, 274
univariate, 276
update_weights_strat_wilson, 277
update_weights_strat_wilson(), 197
utils_split_funs, 278
viewport. 83
viewport(), 27, 94, 251