# Package ‘expss’

**November 15, 2020**

**Type** Package  

**Title** Tables, Labels and Some Useful Functions from Spreadsheets and 'SPSS' Statistics  

**Version** 0.10.7  

**Maintainer** Gregory Demin <gdemin@gmail.com>  

**URL** https://gdemin.github.io/expss/  

**BugReports** https://github.com/gdemin/expss/issues  

**Depends** R (>= 3.3.0),  

**Imports** foreign, utils, stats, magrittr (>= 1.5), htmlTable (>= 1.11.0), matrixStats (>= 0.51.0), data.table (>= 1.10),  

**Suggests** DT, htmltools, knitr, rmarkdown, repr, ggplot2, testthat, openxlsx, fst, huxtable  

**Description** Package computes and displays tables with support for 'SPSS'-style labels, multiple and nested banners, weights, multiple-response variables and significance testing. There are facilities for nice output of tables in 'knitr', 'Shiny', '*.xlsx' files, R and 'Jupyter' notebooks. Methods for labelled variables add value labels support to base R functions and to some functions from other packages. Additionally, the package brings popular data transformation functions from 'SPSS' Statistics and 'Excel': 'RECODE', 'COUNT', 'COMPUTE', 'DO IF', 'COUNTIF', 'VLOOKUP' and etc. These functions are very useful for data processing in marketing research surveys. Package intended to help people to move data processing from 'Excel' and 'SPSS' to R.  

**VignetteBuilder** knitr  

**LazyData** yes  

**License** GPL (>= 2)  

**RoxygenNote** 7.1.1  

**NeedsCompilation** no  

**Author** Gregory Demin [aut, cre],  

Sebastian Jeworutzki [ctb] (<https://orcid.org/0000-0002-2671-5253>)  

**Repository** CRAN  

**Date/Publication** 2020-11-15 22:00:06 UTC
R topics documented:

<table>
<thead>
<tr>
<th>R topics documented:</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>add_rows</td>
<td>3</td>
</tr>
<tr>
<td>apply_labels</td>
<td>4</td>
</tr>
<tr>
<td>as.category</td>
<td>5</td>
</tr>
<tr>
<td>asdatatable_widget</td>
<td>7</td>
</tr>
<tr>
<td>as.dichotomy</td>
<td>8</td>
</tr>
<tr>
<td>as.etable</td>
<td>11</td>
</tr>
<tr>
<td>as.labeled</td>
<td>12</td>
</tr>
<tr>
<td>as_huxtable.etable</td>
<td>13</td>
</tr>
<tr>
<td>compare_proportions</td>
<td>14</td>
</tr>
<tr>
<td>compute</td>
<td>16</td>
</tr>
<tr>
<td>count_if</td>
<td>19</td>
</tr>
<tr>
<td>criteria</td>
<td>25</td>
</tr>
<tr>
<td>cro</td>
<td>29</td>
</tr>
<tr>
<td>cro_fun</td>
<td>35</td>
</tr>
<tr>
<td>default_dataset</td>
<td>42</td>
</tr>
<tr>
<td>do_repeat</td>
<td>43</td>
</tr>
<tr>
<td>drop_empty_rows</td>
<td>45</td>
</tr>
<tr>
<td>experimental</td>
<td>46</td>
</tr>
<tr>
<td>expss</td>
<td>49</td>
</tr>
<tr>
<td>expss.options</td>
<td>51</td>
</tr>
<tr>
<td>fctr</td>
<td>53</td>
</tr>
<tr>
<td>fre</td>
<td>54</td>
</tr>
<tr>
<td>htmlTable.etable</td>
<td>56</td>
</tr>
<tr>
<td>if_na</td>
<td>60</td>
</tr>
<tr>
<td>info</td>
<td>62</td>
</tr>
<tr>
<td>keep</td>
<td>63</td>
</tr>
<tr>
<td>match_row</td>
<td>64</td>
</tr>
<tr>
<td>merge.etable</td>
<td>66</td>
</tr>
<tr>
<td>mrset</td>
<td>67</td>
</tr>
<tr>
<td>names2labels</td>
<td>68</td>
</tr>
<tr>
<td>name_dots</td>
<td>69</td>
</tr>
<tr>
<td>nest</td>
<td>70</td>
</tr>
<tr>
<td>net</td>
<td>71</td>
</tr>
<tr>
<td>prepend_values</td>
<td>66</td>
</tr>
<tr>
<td>product_test</td>
<td>75</td>
</tr>
<tr>
<td>prop</td>
<td>76</td>
</tr>
<tr>
<td>qe</td>
<td>77</td>
</tr>
<tr>
<td>read_spss</td>
<td>78</td>
</tr>
<tr>
<td>recode</td>
<td>79</td>
</tr>
<tr>
<td>ref</td>
<td>80</td>
</tr>
<tr>
<td>set_caption</td>
<td>84</td>
</tr>
<tr>
<td>sheet</td>
<td>86</td>
</tr>
<tr>
<td>sort_asc</td>
<td>87</td>
</tr>
<tr>
<td>split_by</td>
<td>88</td>
</tr>
<tr>
<td>split_labels</td>
<td>89</td>
</tr>
<tr>
<td>split_labels</td>
<td>90</td>
</tr>
<tr>
<td>sum_row</td>
<td>92</td>
</tr>
</tbody>
</table>
**add_rows**

Add rows to data.frame/matrix/table

---

**Description**

Take a sequence of vector, matrix or data-frame arguments and combine by rows. Contrary to `rbind` it handles non-matching column names. `%add_rows%` is an infix version of `add_rows`. There is also special method for the results of `cro/cro_fun/tables/fre`. `.add_rows` is version for adding rows to default dataset. See `default_dataset`.

**Usage**

```r
add_rows(...)  
```

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

```r
add_rows(..., nomatch_columns = c("add", "drop", "stop"))
```

```r
x %add_rows% y
```

```r
.add_rows(..., nomatch_columns = c("add", "drop", "stop"))
```

**Arguments**

- `...` data.frame/matrix/table for binding
- `nomatch_columns` action if there are non-matching columns between data.frames. Possible values are "add", "drop", "stop". "add" will combine all columns, "drop" will leave only common columns, "stop" will raise an error.
- `x` data.frame/matrix/table for binding
- `y` data.frame/matrix/table for binding
**Value**

See `rbind, cro, cro_fun, fre, tables`

**Examples**

```r
a = data.frame(x = 1:5, y = 6:10)
b = data.frame(y = 6:10, z = 11:15)

add_rows(a, b) # x, y, z
a %add_rows% b # the same result

add_rows(a, b, nomatch_columns = "drop") # y

# simple tables
data(mtcars)
# apply labels
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (lb/1000)",
  qsec = "1/4 mile time",
  vs = "V/S",
  am = c("V-engine" = 0, "Straight engine" = 1),
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

tbl_mean = calculate(mtcars, cro_mean(cyl, am))
tbl_percent = calculate(mtcars, cro_cpct(cyl, am))

tbl_mean %add_rows% tbl_percent
```

---

**Description**

`apply_labels` tries automatically detect what is variable label and what are value labels. `.apply_labels` is version for working with default dataset. See also `var_lab` and `val_lab`.

**Usage**

```r
apply_labels(data, ...)

.apply_labels(...)```
as.category

Arguments

data data.frame/list
... named arguments. Name of argument is a variable name in data. Argument values are variable or value labels. Unnamed characters of length 1 are considered as variable labels and named vectors are considered as value labels.

Value
data with applied labels

Examples
data(mtcars)
mtcars = apply_labels(mtcars,
  vs = "Engine",
  vs = num_lab("0 V-engine
1 Straight engine
"),
  am = "Transmission",
  am = num_lab("0 Automatic
1 Manual
")
)

# 'table' from base R
table(mtcars$vs, mtcars$am)

# more sophisticated crosstable
calculate(mtcars, cro(vs, am))

---

as.category Convert dichotomy data.frame/matrix to data.frame with category encoding

Description

Convert dichotomy data.frame/matrix to data.frame with category encoding

Usage

as.category(x, prefix = NULL, counted_value = 1, compress = FALSE)

is.category(x)
Arguments

- **x**: Dichotomy data.frame/matrix (usually with 0,1 coding).
- **prefix**: If is not NULL then column names will be added in the form prefix+column number.
- **counted_value**: Vector. Values that will be considered as indicator of category presence. By default it equals to 1.
- **compress**: Logical. Should we drop columns with all NA? FALSE by default. TRUE significantly decreases performance of the function.

Value
data.frame of class `category` with numeric values that correspond to column numbers of counted values. Column names of `x` or variable labels are added as value labels.

See Also

- `as.dichotomy` for reverse conversion, `mrset`, `mdset` for usage multiple-response variables with tables.

Examples

```r
set.seed(123)

# Let's imagine it's matrix of consumed products
dichotomy_matrix = matrix(sample(c(0,1),40,replace = TRUE,prob=c(.6,.4)),nrow=10)
colnames(dichotomy_matrix) = c("Milk","Sugar","Tea","Coffee")

as.category(dichotomy_matrix, compress = TRUE) # compressed version
category_encoding = as.category(dichotomy_matrix)

# should be TRUE
identical(val_lab(category_encoding), c(Milk = 1L, Sugar = 2L, Tea = 3L, Coffee = 4L))
all(as.dichotomy(category_encoding, use_na = FALSE) == dichotomy_matrix)

# with prefix
as.category(dichotomy_matrix, prefix = "products_")

# data.frame with variable labels
dichotomy_dataframe = as.data.frame(dichotomy_matrix)
colnames(dichotomy_dataframe) = paste0("product_", 1:4)
var_lab(dichotomy_dataframe[[1]]) = "Milk"
var_lab(dichotomy_dataframe[[2]]) = "Sugar"
var_lab(dichotomy_dataframe[[3]]) = "Tea"
var_lab(dichotomy_dataframe[[4]]) = "Coffee"

as.category(dichotomy_dataframe, prefix = "products_")
```
Create an HTML table widget for usage with Shiny

Description

This is method for rendering results of tables/fire/cro in Shiny. DT package should be installed for this feature (install.packages('DT')). For detailed description of function and its arguments see \texttt{datatable}.

Usage

\begin{verbatim}
asdatatable_widget(data, ...) \\
\# S3 method for class 'etable'
asdatatable_widget(
  data,
  ..., 
  repeat_row_labels = FALSE,
  show_row_numbers = FALSE,
  digits = get_expss_digits()
)
\end{verbatim}

\begin{verbatim}
\# S3 method for class 'with_caption'
asdatatable_widget(
  data,
  ..., 
  repeat_row_labels = FALSE,
  show_row_numbers = FALSE,
  digits = get_expss_digits()
)
\end{verbatim}

Arguments

\begin{itemize}
  \item \texttt{data} a data object (result of tables/fire/cro).
  \item \texttt{...} further parameters for \texttt{datatable}
  \item \texttt{repeat_row_labels} logical Should we repeat duplicated row labels in the every row? Default is FALSE.
  \item \texttt{show_row_numbers} logical Default is FALSE.
  \item \texttt{digits} integer By default, all numeric columns are rounded to one digit after decimal separator. Also you can set this argument by option 'expss.digits' - for example, \texttt{expss_digits(2)}. If it is NA than all numeric columns remain unrounded.
\end{itemize}

Value

Object of class \texttt{datatable}
## not run:

data(mtcars)
mtcars = apply_labels(mtcars, 
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (1000 lbs)",
  qsec = "1/4 mile time",
  vs = "Engine",
  am = "Transmission",
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

library(shiny)
shinyApp(
  ui = fluidPage(fluidRow(column(12, DT::dataTableOutput('tbl')))),
  server = function(input, output) {
    output$tbl = DT::renderDataTable(asdatatable_widget(mtcars_table)
  )
  }
)

## End(Not run)

as.dichotomy

Convert variable (possibly multiple choice question) to
data.frame/matrix of dummy variables.
Description

This function converts variable/multiple response variable (vector/matrix/data.frame) with category encoding into data.frame/matrix with dichotomy encoding (0/1) suited for most statistical analysis, e.g. clustering, factor analysis, linear regression and so on.

- `as.dichotomy` returns data.frame of class ‘dichotomy’ with 0, 1 and possibly NA.
- `dummy` returns matrix of class ‘dichotomy’ with 0, 1 and possibly NA.
- `dummy1` drops last column in dichotomy matrix. It is useful in many cases because any column of such matrix usually is linear combinations of other columns.

Usage

```
as.dichotomy(  
  x,    
  prefix = "v",   
  keep_unused = FALSE, 
  use_na = TRUE,  
  keep_values = NULL, 
  keep_labels = NULL, 
  drop_values = NULL, 
  drop_labels = NULL,  
  presence = 1, 
  absence = 0   
)

dummy(  
  x,    
  keep_unused = FALSE, 
  use_na = TRUE,  
  keep_values = NULL, 
  keep_labels = NULL, 
  drop_values = NULL, 
  drop_labels = NULL,  
  presence = 1, 
  absence = 0   
)

dummy1(  
  x,    
  keep_unused = FALSE, 
  use_na = TRUE,  
  keep_values = NULL, 
  keep_labels = NULL, 
  drop_values = NULL, 
  drop_labels = NULL,  
  presence = 1, 
  absence = 0  
)
```
is.dichotomy(x)

Arguments

x vector/factor/matrix/data.frame.
prefix character. By default "v".
keep_unused Logical. Should we create columns for unused value labels/factor levels? FALSE by default.
use_na Logical. Should we use NA for rows with all NA or use 0's instead. TRUE by default.
keep_values Numeric/character. Values that should be kept. By default all values will be kept.
keep_labels Numeric/character. Labels/levels that should be kept. By default all labels/levels will be kept.
drop_values Numeric/character. Values that should be dropped. By default all values will be kept. Ignored if keep_values/keep_labels are provided.
drop_labels Numeric/character. Labels/levels that should be dropped. By default all labels/levels will be kept. Ignored if keep_values/keep_labels are provided.
presence numeric value which will code presence of the level. By default it is 1. Note that all tables functions need that presence and absence will be 1 and 0.
score numeric value which will code absence of the level. By default it is 0. Note that all tables functions need that presence and absence will be 1 and 0.

Value

as.dichotomy returns data.frame of class dichotomy with 0,1. Columns of this data.frame have variable labels according to value labels of original data. If label doesn’t exist for particular value then this value will be used as variable label. dummy returns matrix of class dichotomy. Column names of this matrix are value labels of original data.

See Also

as.category for reverse conversion, mrset, mdset for usage multiple-response variables with tables.

Examples

# toy example
# brands - multiple response question
# Which brands do you use during last three months?
set.seed(123)
brands = as.sheet(t(replicate(20,sample(c(1:5,NA),4,replace = FALSE))))
# score - evaluation of tested product
score = sample(-1:1,20,replace = TRUE)
var_lab(brands) = "Used brands"
val_lab(brands) = autonum("
Brand A
Brand B
Brand C
Brand D
Brand E
"

var_lab(score) = "Evaluation of tested brand"
val_lab(score) = make_labels("-1 Dislike it
   0 So-so
   1 Like it"
"

cro_cpct(as.dichotomy(brands), score)
# the same as
cro_cpct(mrset(brands), score)

# customer segmentation by used brands
kmeans(dummy(brands), 3)

# model of influence of used brands on evaluation of tested product
summary(lm(score ~ dummy(brands)))

# prefixed data.frame
as.dichotomy(brands, prefix = "brand_")

---

as.etable  

Convert data.frame/matrix to object of class ‘etable’

Description

If `x` is `data.frame` then `as.etable` just adds `etable` to class attribute of `x`. If `x` is matrix then it will be converted to `data.frame`.

Usage

```r
as.etable(x, rownames_as_row_labels = NULL)
```

```r
is.etable(x)
```

Arguments

- `x`  
  data.frame/matrix
- `rownames_as_row_labels`  
  logical. If it is TRUE than rownames of `x` will be added to result as first column with name `row_labels`. By default row names will be added if they are not NULL and are not sequential numerics.
Value

object of class etable

Examples

data(mtcars)
etable_mtcars = as.etable(mtcars)
is.etable(table_mtcars) #TRUE

etable_mtcars #another 'print' method is used
cor(mtcars) %>% as.etable()

---

**as.labelled**

Recode vector into numeric vector with value labels

Description

Recode vector into numeric vector with value labels

Usage

as.labelled(x, label = NULL)

is.labelled(x)

Arguments

x numeric vector/character vector/factor

label optional variable label

Value

numeric vector with value labels

Examples

character_vector = c("one", "two", "two", "three")
as.labelled(character_vector, label = "Numbers")
as_huxtable.etable    Convert table to huxtable

Description
This function converts a etable object to a huxtable. The huxtable-package needs to be installed to use this function.

Usage
as_huxtable.etable(x, ...)
as_hux.etable(x, ...)

Arguments
x    etable. Table to convert to a huxtable.
...  arguments passed on to huxtable.

Details
huxtable allows to export formatted tables to LaTeX, HTML, Microsoft Word, Microsoft Excel, Microsoft Powerpoint, RTF and Markdown.

Tables in knitr or rmarkdown documents of type LaTeX or Word are converted by default.

Examples
## Not run:
library(huxtable)
data(mtcars)
mtcars = apply_labels(mtcars,
    mpg = "Miles/(US) gallon",
cyl = "Number of cylinders",
disp = "Displacement (cu.in.)",
hp = "Gross horsepower",
drat = "Rear axle ratio",
wt = "Weight (1000 lbs)",
qsec = "1/4 mile time",
vs = "Engine",
vs = c("V-engine" = 0,
    "Straight engine" = 1),
am = "Transmission",
am = c("Automatic" = 0,
    "Manual"=1),
gear = "Number of forward gears",
carb = "Number of carburetors"
)

tab = mtcars %>%
compare_proportions

Description

compare_proportions calculates p-values (via z-test) for comparison between each proportion in the prop1 and prop2. Results are calculated with the same formula as in prop.test without continuity correction. compare_means calculates p-values (via t-test) for comparison between each mean in the mean1 and mean2. Results are calculated on the aggregated statistics (means, std. devs, N) with the same formula as in t.test. These functions mainly intended for usage inside significance_cpct and significance_means.

Usage

compare_proportions(prop1, prop2, base1, base2, common_base = 0)

compare_means(
  mean1,
  mean2,
  sd1,
  sd2,
  base1,
  base2,
  common_base = 0,
  var_equal = FALSE
)

Arguments

prop1 a numeric vector of proportions in the group 1. Values should be between 0 and 1
prop2 a numeric vector of proportions in the group 2. Values should be between 0 and 1
base1  a numeric vector for compare_means and single number for compare_proportions. Number of valid cases for each mean in the first group for compare_means and number of cases for compare_proportions.

base2  a numeric vector for compare_means and single number for compare_proportions. Number of valid cases for each mean in the second group for compare_means and number of cases for compare_proportions.

common_base  numeric. Number of cases that belong to both values in the first and the second argument. It can occur in the case of overlapping samples. Calculations are made according to algorithm in IBM SPSS Statistics Algorithms v20, p. 263. Note that with these adjustments t-tests between means are made with equal variance assumed (as with var_equal = TRUE).

mean1  a numeric vector of the means in the first group.

mean2  a numeric vector of the means in the second group.

sd1  a numeric vector of the standard deviations in the first group. Values should be non-negative.

sd2  a numeric vector of the standard deviations in the second group. Values should be non-negative.

var_equal  a logical variable indicating whether to treat the variances in the groups as being equal. For details see t.test.

Value  numeric vector with p-values

See Also  significance_cpcct, significance_means, prop.test, t.test

Examples

# proportions
data(mtcars)
counts = table(mtcars$am, mtcars$vs)
props = prop.table(counts)
compare_proportions(props[,1], props[,2],
                   colSums(counts)[1], colSums(counts)[1])

# means
t.test(mpg ~ am, data = mtcars)$p.value
# the same result
calculate(mtcars,
          compare_means(
            mean(mpg[am==0]), mean(mpg[am==1]),
            sd(mpg[am==0]), sd(mpg[am==1]),
            length(mpg[am==0]), length(mpg[am==1])
          ))
compute Modify data.frame/modify subset of the data.frame

Description

- **compute** evaluates expression **expr** in the context of data.frame **data** and return original data possibly modified.

- **calculate** evaluates expression **expr** in the context of data.frame **data** and return value of the evaluated expression. Function **use_labels** is shortcut for **calculate** with argument **use_labels** set to **TRUE**. When **use_labels** is **TRUE** there is a special shortcut for entire data.frame - **..data**.

- **do_if** modifies only rows for which **cond** equals to **TRUE**. Other rows remain unchanged. Newly created variables also will have values only in rows for which **cond** have **TRUE**. There will be NA's in other rows. This function tries to mimic SPSS "DO IF(). ... END IF." statement.

Full-featured %to% is available in the expressions for addressing range of variables. There is a special constant .N which equals to number of cases in **data** for usage in expression inside **compute/calculate**. Inside **do_if** .N gives number of rows which will be affected by expressions. For parametrization (variable substitution) see .. or examples. Sometimes it is useful to create new empty variable inside compute. You can use .new_var function for this task. This function creates variable of length .N filled with NA. See examples. **modify** is an alias for **compute**, **modify_if** is an alias for **do_if** and **calc** is an alias for **calculate**.

Usage

```r
compute(data, ...)  
modify(data, ...)  
do_if(data, cond, ...)  
modify_if(data, cond, ...)  
calculate(data, expr, use_labels = FALSE)  
use_labels(data, expr)  
calc(data, expr, use_labels = FALSE)  
data %calc% expr  
data %use_labels% expr  
data %calculate% expr
```

Arguments

data data.frame/list of data.frames. If data is list of data.frames then expression expr will be evaluated inside each data.frame separately.

... expressions that should be evaluated in the context of data.frame data. It can be arbitrary code in curly brackets or assignments. See examples.

cond logical vector or expression. Expression will be evaluated in the context of the data.

expr expression that should be evaluated in the context of data.frame data

use_labels logical. Experimental feature. If it equals to TRUE then we will try to replace variable names with labels. So many base R functions which show variable names will show labels.

Value

compute and do_if functions return modified data.frame/list of modified data.frames, calculate returns value of the evaluated expression/list of values.

Examples

dfs = data.frame(
  test = 1:5,
  a = rep(10, 5),
  b_1 = rep(11, 5),
  b_2 = rep(12, 5),
  b_3 = rep(13, 5),
  b_4 = rep(14, 5),
  b_5 = rep(15, 5)
)

# compute sum of b* variables and attach it to 'dfs'
compute(dfs, {
  b_total = sum_row(b_1 %to% b_5)
  var_lab(b_total) = "Sum of b"
  random_numbers = runif(.N) # .N usage
})

# calculate sum of b* variables and return it
calculate(dfs, sum_row(b_1 %to% b_5))

# set values to existing/new variables
compute(dfs, {
  (b_1 %to% b_5) %into% text_expand('new_b{1:5}')
})

# .new_var usage
compute(dfs, {
  new_var = .new_var()
  new_var[1] = 1 # this is not possible without preliminary variable creation
})
# conditional modification
do_if(dfs, test %in% 2:4, {
  a = a + 1
  b_total = sum_row(b_1 %to% b_5)
  random_numbers = runif(.N) # .N usage
})

# variable substitution
name1 = "a"
name2 = "new_var"
compute(dfs, {
  ..$name2 = ..$name1*2
})

compute(dfs, {
  for(name1 in paste0("b_", 1:5)){
    name2 = paste0("new_", name1)
    ..$name2 = ..$name1*2
  }
  rm(name1, name2) # we don't need this variables as columns in 'dfs'
})

# square brackets notation
compute(dfs, {
  ..[(name2)] = ..[(name1)]*2
})

compute(dfs, {
  for(name1 in paste0("b_", 1:5)){
    ..[paste0("new_", name1)] = ..$name1*2
  }
  rm(name1) # we don't need this variable as column in 'dfs'
})

# '..$' doesn't work for case below so we need to use square brackets form
name1 = paste0("b_", 1:5)
name2 = paste0("new_", name1)
compute(dfs, {
  for(i in 1:5){
    ..[name2[i]] = ..[name1[i]]*3
  }
  rm(i) # we don't need this variable as column in 'dfs'
})

# 'use_labels' examples. Utilization of labels in base R.
data(mtcars)
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",}
disp = "Displacement (cu.in.)",
hp = "Gross horsepower",
drat = "Rear axle ratio",
wtt = "Weight (lb/1000)",
qsec = "1/4 mile time",
vs = "Engine",
vs = c("V-engine" = 0,
        "Straight engine" = 1),
am = "Transmission",
am = c("Automatic" = 0,
      "Manual" = 1),
gear = "Number of forward gears",
carb = "Number of carburetors"

use_labels(mtcars, table(am, vs))

## Not run:
use_labels(mtcars, plot(mpg, hp))

## End(Not run)

mtcars %>%
  use_labels(lm(mpg ~ disp + hp + wt)) %>%
  summary()

---

**count_if**  
*Count/sum/average/other functions on values that meet a criterion*

**Description**

These functions calculate count/sum/average/etc. on values that meet a criterion that you specify. `apply_if_*` apply custom functions. There are different flavors of these functions: *_if work on entire dataset/matrix/vector, *_row_if works on each row and *_col_if works on each column.

**Usage**

```r
count_if(criterion, ...)
count_row_if(criterion, ...)
count_col_if(criterion, ...)
x %row_in% criterion
x %has% criterion
x %col_in% criterion
```
sum_if(criterion, ..., data = NULL)
sum_row_if(criterion, ..., data = NULL)
sum_col_if(criterion, ..., data = NULL)
mean_if(criterion, ..., data = NULL)
mean_row_if(criterion, ..., data = NULL)
mean_col_if(criterion, ..., data = NULL)
sd_if(criterion, ..., data = NULL)
sd_row_if(criterion, ..., data = NULL)
sd_col_if(criterion, ..., data = NULL)
median_if(criterion, ..., data = NULL)
median_row_if(criterion, ..., data = NULL)
median_col_if(criterion, ..., data = NULL)
max_if(criterion, ..., data = NULL)
max_row_if(criterion, ..., data = NULL)
max_col_if(criterion, ..., data = NULL)
min_if(criterion, ..., data = NULL)
min_row_if(criterion, ..., data = NULL)
min_col_if(criterion, ..., data = NULL)
apply_row_if(fun, criterion, ..., data = NULL)
apply_col_if(fun, criterion, ..., data = NULL)

**Arguments**

- **criterion**: Vector with counted values or function. See details and examples.
- **...**: Data on which criterion will be applied. Vector, matrix, data.frame, list.
- **x**: Data on which criterion will be applied. Vector, matrix, data.frame, list.
- **data**: Data on which function will be applied. Doesn’t applicable to count_*_if functions. If omitted then function will be applied on the ... argument.
fun

Custom function that will be applied based on criterion.

Details

Possible type for criterion argument:

- **vector/single value** All values in ... which equal to the elements of vector in the criteria will be used as function fun argument.
- **function** Values for which function gives TRUE will be used as function fun argument. There are some special functions for convenience (e.g. \texttt{gt(5)} is equivalent ">5" in spreadsheet) - see criteria.

count* and %in% never returns NA's. Other functions remove NA's before calculations (as \texttt{na.rm = TRUE} in base R functions).

Function criterion should return logical vector of same size and shape as its argument. This function will be applied to each column of supplied data and TRUE results will be used. There is asymmetrical behavior in \_,row_if and \_,col_if for function criterion: in both cases function criterion will be applied column-wise.

Value

\_if return single value (vector of length 1). \_row_if returns vector for each row of supplied arguments. \_col_if returns vector for each column of supplied arguments. \%row_in%/\%col_in% return logical vector - indicator of presence of criterion in each row/column. %has% is an alias for %row_in%.

Examples

```r
set.seed(123)
sheet1 = as.sheet(
  matrix(sample(c(1:10,NA), 30, replace = TRUE), 10)
)
result = compute(sheet1, {
  # count 8
  exact = count_row_if(8, V1, V2, V3)
  # count values greater than 8
  greater = count_row_if(gt(8), V1, V2, V3)
  # count integer values between 5 and 8, e.g. 5, 6, 7, 8
  integer_range = count_row_if(5:8, V1, V2, V3)
  # count values between 5 and 8
  range = count_row_if(5 %thru% 8, V1, V2, V3)
  # count NA
  na = count_row_if(is.na, V1, V2, V3)
  # count not-NA
  not_na = count_row_if(not_na, V1, V2, V3)
  # are there any 5 in each row?
  has_five = cbind(V1, V2, V3) %row_in% 5
})
result
```
mean_row_if(6, sheet1$V1, data = sheet1)
median_row_if(gt(2), sheet1$V1, sheet1$V2, sheet1$V3)
sd_row_if(5 %thru% 8, sheet1$V1, sheet1$V2, sheet1$V3)

if_na(sheet1) = 5 # replace NA

# custom apply
apply_col_if(prod, gt(2), sheet1$V1, data = sheet1) # product of all elements by columns
apply_row_if(prod, gt(2), sheet1$V1, data = sheet1) # product of all elements by rows

# Examples borrowed from Microsoft Excel help for COUNTIF
sheet1 = text_to_columns(
    "   a   b
apples   32
oranges   54
peaches   75
apples   86
   "
)
count_if("apples", sheet1$a) # 2
count_if("apples", sheet1) # 2
calc(sheet1, count_if("apples", a, b)) # 2
count_if(gt(55), sheet1$b) # greater than 55 = 2
count_if(ne(75), sheet1$b) # not equal 75 = 3
count_if(ge(32), sheet1$b) # greater than or equal 32 = 4
count_if(gt(32) & lt(86), sheet1$b) # 2
# count only integer values between 33 and 85
count_if(33:85, sheet1$b) # 2
# values with letters
count_if(regex("^[A-z]+$"), sheet1) # 4
# values that started on 'a'
count_if(regex("^a"), sheet1) # 2
# count_row_if
count_row_if(regex("^a"), sheet1) # c(1,0,0,1)
sheet1 %row_in% 'apples' # c(TRUE,FALSE,FALSE,TRUE)

# Some of Microsoft Excel examples for SUMIF/AVERAGEIF/etc
sheet1 = text_to_columns(
    "   property_value commission data"
count_if

calc(sheet1, sum_if(gt(160000), property_value, data = commission)) # 63000

calc(sheet1, sum_if(gt(160000), property_value)) # 900000

calc(sheet1, sum_if(300000, property_value, data = commission)) # 21000

calc(sheet1, sum_if(gt(data[1]), property_value, data = commission)) # 49000

sheet1 = text_to_columns(
  "category food sales
  Vegetables Tomatoes 2300
  Vegetables Celery 5500
  Fruits Oranges 800
  NA Butter 400
  Vegetables Carrots 4200
  Fruits Apples 1200"
)

calc(sheet1, sum_if("Fruits", category, data = sales)) # 2000

calc(sheet1, sum_if("Vegetables", category, data = sales)) # 12000

calc(sheet1, sum_if(perl("es"), food, data = sales)) # 4300

calc(sheet1, sum_if(NA, category, data = sales)) # 400

sheet1 = text_to_columns(
  "property_value commission data
  100000 7000 250000
  200000 14000
  300000 21000
  400000 28000"
)
```plaintext
# Commision average for commission less than 23000
calc(sheet1, mean_if(lt(23000), commission)) # 14000

# Property value average for property value less than 95000
calc(sheet1, mean_if(lt(95000), property_value)) # NaN

# Commision average for property value greater than 250000
calc(sheet1, mean_if(gt(250000), property_value, data = commission)) # 24500

sheet1 = text_to_columns(
  'region profits
    East 45678
    West 23789
    North -4789
    "South (New Office)" 0
    MidWest 9678
  ',
  quote = '"',
)

# Mean profits for 'west' regions
calc(sheet1, mean_if(contains("West"), region, data = profits)) # 16733.5

# Mean profits for regions which doesn't contain New Office
calc(sheet1, mean_if(not(contains("New Office")), region, data = profits)) # 18589

sheet1 = text_to_columns(
  "grade weight
    89 1
    93 2
    96 2
    85 3
    91 1
    88 1
  "
)

# Minimum grade for weight equals to 1
calc(sheet1, min_if(1, weight, data = grade)) # 88

# Maximum grade for weight equals to 1
calc(sheet1, max_if(1, weight, data = grade)) # 91
```
criteria

# Example with offset
sheet1 = text_to_columns(
    
    weight grade
    10  b
    11  a
    100 a
    111 b
    1  a
    1  a

)

calc(sheet1, min_if("a", grade[2:5], data = weight[1:4])) # 10

---

criteria | Criteria functions

---

Description

Produce criteria which could be used in the different situations - see 'recode', 'na_if', 'count_if', 'match_row', '%i%' and etc. For example, 'greater(5)' returns function which tests whether its argument greater than five. 'fixed("apple")' returns function which tests whether its argument contains "apple". For criteria logical operations (|, &, !, xor) are defined, e. g. you can write something like: 'greater(5) | equals(1)'. List of functions:

- comparison criteria - 'equals', 'greater' and etc. return functions which compare its argument against value.
- 'thru' checks whether a value is inside interval. 'thru(0,1)' is equivalent to 'x>=0 & x<=1'
- '%thru%' is infix version of 'thru', e. g. '0 %thru% 1'
- 'is_max' and 'is_min' return TRUE where vector value is equals to maximum or minimum.
- 'contains' searches for the pattern in the strings. By default, it works with fixed patterns rather than regular expressions. For details about its arguments see grepl
- 'like' searches for the Excel-style pattern in the strings. You can use wildcards: '*' means any number of symbols, '?' means single symbol. Case insensitive.
- 'fixed' alias for contains.
- 'perl' such as 'contains' but the pattern is perl-compatible regular expression ('perl = TRUE'). For details see grepl
- 'regex' use POSIX 1003.2 extended regular expressions ('fixed = FALSE'). For details see grepl
- 'has_label' searches values which have supplied label(-s). We can used criteria as an argument for 'has_label'.
- 'to' returns function which gives TRUE for all elements of vector before the first occurrence of 'x' and for 'x'.
• 'from' returns function which gives TRUE for all elements of vector after the first occurrence of 'x' and for 'x'.
• 'not_na' returns TRUE for all non-NA vector elements.
• 'other' returns TRUE for all vector elements. It is intended for usage with 'recode'.
• 'items' returns TRUE for the vector elements with the given sequential numbers.
• 'and', 'or', 'not' are spreadsheet-style boolean functions.

Shortcuts for comparison criteria:
• 'equals' - 'eq'
• 'not_equals' - 'neq', 'ne'
• 'greater' - 'gt'
• 'greater_or_equal' - 'gte', 'ge'
• 'less' - 'lt'
• 'less_or_equal' - 'lte', 'le'

Usage

as.criterion(crit)

is.criterion(x)

equals(x)

not_equals(x)

less(x)

less_or_equal(x)

greater(x)

greater_or_equal(x)

thru(lower, upper)

lower %thru% upper

when(x)

is_max(x)

is_min(x)

contains(
    pattern,
    ignore.case = FALSE,
perl = FALSE,  
fixed = TRUE,  
useBytes = FALSE)
)

like(pattern)

fixed(
    pattern,  
    ignore.case = FALSE,  
    perl = FALSE,  
    fixed = TRUE,  
    useBytes = FALSE)
)

perl(
    pattern,  
    ignore.case = FALSE,  
    perl = TRUE,  
    fixed = FALSE,  
    useBytes = FALSE)
)

regex(
    pattern,  
    ignore.case = FALSE,  
    perl = FALSE,  
    fixed = FALSE,  
    useBytes = FALSE)
)

has_label(x)

from(x)

to(x)

items(...)  

not_na(x)

is_na(x)

other(x)

and(...)  
or(...)
not(x)

Arguments

crit vector of values/function which returns logical or logical vector. It will be converted to function of class criterion.
x vector
lower vector/single value - lower bound of interval
upper vector/single value - upper bound of interval
pattern character string containing a regular expression (or character string for 'fixed') to be matched in the given character vector. Coerced by as.character to a character string if possible.
ignore.case logical see grepl
perl logical see grepl
fixed logical see grepl
useBytes logical see grepl
... numeric indexes of desired items for items, logical vectors or criteria for boolean functions.

Value

function of class 'criterion' which tests its argument against condition and return logical value

See Also

recode, count_if, match_row, na_if, %i%

Examples

# operations on vector, '%d%' means 'diff'
1:6 %d% greater(4) # 1:4
1:6 %d% (1 | greater(4)) # 2:4
# '%i%' means 'intersect'
1:6 %i% (is_min() | is_max()) # 1, 6
# with Excel-style boolean operators
1:6 %i% or(is_min(), is_max()) # 1, 6

letters %i% (contains("a") | contains("z")) # a, z
letters %i% perl("a|z") # a, z
letters %i% from("w") # w, x, y, z
letters %i% to("c") # a, b, c
letters %i% (from("b") & to("e")) # b, d, e
c(1, 2, NA, 3) %>% not_na() # (1, 2)

# examples with count_if
df1 = data.frame(
  a=c("apples", "oranges", "peaches", "apples"),
  b = c(32, 54, 75, 86)
)

count_if(greater(55), df1$b) # greater than 55 = 2
count_if(not_equals(75), df1$b) # not equals 75 = 3

count_if(greater(32) & less(86), df1$b) # greater than 32 and less than 86 = 2
count_if(and(greater(32), less(86)), df1$b) # the same result

# infix version
count_if(35 %thru% 80, df1$b) # greater than or equals to 35 and less than or equals to 80 = 2

# values that started on 'a'
count_if(like("a"), df1) # 2

# the same with Perl-style regular expression
count_if(perl("^a"), df1) # 2

# count_row_if
count_row_if(perl("^a"), df1) # c(1,0,0,1)

# examples with 'n_intersect' and 'n_diff'
data(iris)
iris %>% n_intersect(to("Petal.Width")) # all columns up to 'Species'

# 'Sepal.Length', 'Sepal.Width' will be left
iris %>% n_diff(from("Petal.Length"))

# except first column
iris %n_d% items(1)

# 'recode' examples
qvar = c(1:20, 97, NA, NA)
recode(qvar, 1 %thru% 5 ~ 1, 6 %thru% 10 ~ 2, 11 %thru% hi ~ 3, other ~ 0)

# the same result
recode(qvar, 1 %thru% 5 ~ 1, 6 %thru% 10 ~ 2, greater_or_equal(11) ~ 3, other ~ 0)

cro 

---

cro

Cross tabulation with support of labels, weights and multiple response variables.
Description

- cro, cro_cases build a contingency table of the counts.
- cro_cpct, cro_cpct_responses build a contingency table of the column percent. These functions give different results only for multiple response variables. For cro_cpct base of percent is number of valid cases. Case is considered as valid if it has at least one non-NA value. So for multiple response variables sum of percent may be greater than 100. For cro_cpct_responses base of percent is number of valid responses. Multiple response variables can have several responses for single case. Sum of percent of cro_cpct_responses always equals to 100%.
- cro_rpct build a contingency table of the row percent. Base for percent is number of valid cases.
- cro_tpct build a contingency table of the table percent. Base for percent is number of valid cases.
- calc_cro_* are the same as above but evaluate their arguments in the context of the first argument data.

• total auxiliary function - creates variables with 1 for valid case of its argument x and NA in opposite case.

You can combine tables with add_rows and merge.etable. For sorting table see tab_sort_asc. To provide multiple-response variables as arguments use mrset for multiples with category encoding and mdset for multiples with dichotomy (dummy) encoding. To compute statistics with nested variables/banners use nest. For more sophisticated interface with modern piping via magrittr see tables.

Usage

cro(
    cell_vars,
    col_vars = total(),
    row_vars = NULL,
    weight = NULL,
    subgroup = NULL,
    total_label = NULL,
    total_statistic = "u_cases",
    total_row_position = c("below", "above", "none")
)

cro_cases(
    cell_vars,
    col_vars = total(),
    row_vars = NULL,
    weight = NULL,
    subgroup = NULL,
    total_label = NULL,
    total_statistic = "u_cases",
    total_row_position = c("below", "above", "none")
)
cro.cpct(
  cell_vars,
  col_vars = total(),
  row_vars = NULL,
  weight = NULL,
  subgroup = NULL,
  total_label = NULL,
  total_statistic = "u_cases",
  total_row_position = c("below", "above", "none")
)

cro.rpct(
  cell_vars,
  col_vars = total(),
  row_vars = NULL,
  weight = NULL,
  subgroup = NULL,
  total_label = NULL,
  total_statistic = "u_cases",
  total_row_position = c("below", "above", "none")
)

cro.tpct(
  cell_vars,
  col_vars = total(),
  row_vars = NULL,
  weight = NULL,
  subgroup = NULL,
  total_label = NULL,
  total_statistic = "u_cases",
  total_row_position = c("below", "above", "none")
)

cro.cpct_responses(
  cell_vars,
  col_vars = total(),
  row_vars = NULL,
  weight = NULL,
  subgroup = NULL,
  total_label = NULL,
  total_statistic = "u_responses",
  total_row_position = c("below", "above", "none")
)

calc_cro(
  data,
  cell_vars,

col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_cases",
total_row_position = c("below", "above", "none")
)

calc_cro_cases(
data,
cell_vars,
col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_cases",
total_row_position = c("below", "above", "none")
)

calc_cro_cpct(
data,
cell_vars,
col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_cases",
total_row_position = c("below", "above", "none")
)

calc_cro_rpct(
data,
cell_vars,
col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_cases",
total_row_position = c("below", "above", "none")
)

calc_cro_tpct(
data,
cell_vars,
col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_cases",
total_row_position = c("below", "above", "none")
)

calc_cro_cpct_responses(
data,
cell_vars,
col_vars = total(),
row_vars = NULL,
weight = NULL,
subgroup = NULL,
total_label = NULL,
total_statistic = "u_responses",
total_row_position = c("below", "above", "none")
)

total(x = 1, label = "#Total")

Arguments

cell_vars vector/data.frame/list. Variables on which percentage/cases will be computed. Use mrset/mdset for multiple-response variables.

col_vars vector/data.frame/list. Variables which breaks table by columns. Use mrset/mdset for multiple-response variables.

row_vars vector/data.frame/list. Variables which breaks table by rows. Use mrset/mdset for multiple-response variables.

weight numeric vector. Optional cases weights. Cases with NA's, negative and zero weights are removed before calculations.

subgroup logical vector. You can specify subgroup on which table will be computed.

total_label By default "#Total". You can provide several names - each name for each total statistics.

total_statistic By default it is "u_cases" (unweighted cases). Possible values are "u_cases", "u_responses", "u_cpct", "u_rpct", "u_tpct", "w_cases", "w_responses", "w_cpct", "w_rpct", "w_tpct". "u_" means unweighted statistics and "w_" means weighted statistics.

total_row_position Position of total row in the resulting table. Can be one of "below", "above", "none".

data data.frame in which context all other arguments will be evaluated (for calc_cro_*)

x vector/data.frame of class 'category'/'dichotomy'.

label character. Label for total variable.
Value

object of class 'etable'. Basically it’s a data.frame but class is needed for custom methods.

See Also

tables, fre, cro_fun.

Examples

data(mtcars)
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (1000 lbs)",
  qsec = "1/4 mile time",
  vs = "Engine",
  vs = c("V-engine" = 0,
      "Straight engine" = 1),
  am = "Transmission",
  am = c("Automatic" = 0,
      "Manual"=1),
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

calculate(mtcars, cro(am, vs))
calc_cro(mtcars, am, vs) # the same result

# column percent with multiple banners
calculate(mtcars, cro_cpct(cyl, list(total(), vs, am)))
calc_cro_cpct(mtcars, cyl, list(total(), vs, am)) # the same result

# nested banner
calculate(mtcars, cro_cpct(cyl, list(total(), vs %nest% am)))

# stacked variables
calculate(mtcars, cro(list(cyl, carb), list(total(), vs %nest% am)))

# nested variables
calculate(mtcars, cro_cpct(am %nest% cyl, list(total(), vs)))

# row variables
calculate(mtcars, cro_cpct(cyl, list(total(), vs), row_vars = am))

# several totals above table
calculate(mtcars, cro_cpct(cyl,
  list(total(), vs),
  row_vars = am,
  total_row_position = "above",
  ...
# multiple-choice variable
# brands - multiple response question
# Which brands do you use during last three months?
set.seed(123)
brands = data.frame(t(replicate(20,sample(c(1:5,NA),4,replace = FALSE))))
# score - evaluation of tested product
score = sample(-1:1,20,replace = TRUE)

var_lab(brands) = "Used brands"
val_lab(brands) = make_labels("1 Brand A
2 Brand B
3 Brand C
4 Brand D
5 Brand E"
)

var_lab(score) = "Evaluation of tested brand"
val_lab(score) = num_lab("-1 Dislike it
0 So-so
1 Like it"
)

cro_cpct(mrset(brands), list(total(), score))
# responses
cro_cpct_responses(mrset(brands), list(total(), score))

cro_fun

Cross-tabulation with custom summary function.

Description

- `cro_mean, cro_sum, cro_median` calculate mean/sum/median by groups. NA's are always omitted.

- `cro_mean_sd_n` calculates mean, standard deviation and N simultaneously. Mainly intended for usage with `significance_means`.

- `cro_pearson, cro_spearman` calculate correlation of first variable in each data.frame in `cell_vars` with other variables. NA's are removed pairwise.

- `cro_fun, cro_fun_df` return table with custom summary statistics defined by `fun` argument. NA's treatment depends on your `fun` behavior. To use weight you should have formal weight argument in `fun` and some logic for its processing inside. Several functions with weight support are provided - see `w_mean`. `cro_fun` applies `fun` on each variable in `cell_vars` separately, `cro_fun_df` gives to `fun` each data.frame in `cell_vars` as a whole. So `cro_fun(iris[, -5], iris$Species, fun = mean)` gives the same result as `cro_fun_df(iris[, -5], iris$Species, fun = mean)`.

- `cro_fun` applies `fun` on each variable in `cell_vars` separately, `cro_fun_df` gives to `fun` each data.frame in `cell_vars` as a whole. So `cro_fun(iris[, -5], iris$Species, fun = mean)` gives the same result as `cro_fun_df(iris[, -5], iris$Species, fun = mean)`. 

- `cro_fun` applies `fun` on each variable in `cell_vars` separately, `cro_fun_df` gives to `fun` each data.frame in `cell_vars` as a whole. So `cro_fun(iris[, -5], iris$Species, fun = mean)` gives the same result as `cro_fun_df(iris[, -5], iris$Species, fun = mean)`.
cro_fun

= colMeans). For cro_fun_df names of cell_vars will converted to labels if they are avail-
able before the fun will be applied. Generally it is recommended that fun will always return
object of the same form. Row names/vector names of fun result will appear in the row labels
of the table and column names/names of list will appear in the column labels. If your fun
returns data.frame/matrix/list with element named 'row_labels' then this element will be used
as row labels. And it will have precedence over rownames.

• calc_cro_* are the same as above but evaluate their arguments in the context of the first
argument data.

• combine_functions is auxiliary function for combining several functions into one func-
tion for usage with cro_fun/cro_fun_df. Names of arguments will be used as statistic
labels. By default, results of each function are combined with c. But you can provide
your own method function with method argument. It will be applied as in the expression
do.call(method,list_of_functions_results). Particular useful method is list. When
it used then statistic labels will appear in the column labels. See examples. Also you may be
interested in data.frame, rbind, cbind methods.

Usage

cro_fun(
  cell_vars,
  col_vars = total(),
  row_vars = total(label = ""),
  weight = NULL,
  subgroup = NULL,
  fun,
  ...,
  unsafe = FALSE
)

cro_fun_df(
  cell_vars,
  col_vars = total(),
  row_vars = total(label = ""),
  weight = NULL,
  subgroup = NULL,
  fun,
  ...,
  unsafe = FALSE
)

cro_mean(
  cell_vars,
  col_vars = total(),
  row_vars = total(label = ""),
  weight = NULL,
  subgroup = NULL
)
cro_fun

cro_mean_sd_n(
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
    subgroup = NULL,
    weighted_valid_n = FALSE,
    labels = NULL
)

cro_sum(
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
    subgroup = NULL
)

cro_median(
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
    subgroup = NULL
)

cro_pearson(
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
    subgroup = NULL
)

cro_spearman(
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
    subgroup = NULL
)

calc_cro_fun(
    data,
    cell_vars,
    col_vars = total(),
    row_vars = total(label = ""),
    weight = NULL,
subgroup = NULL,
fun,
..., unsafe = FALSE
)
calc_cro_fun_df(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL,
fun,
..., unsafe = FALSE
)
calc_cro_mean(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL
)
calc_cro_mean_sd_n(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL,
weighted_valid_n = FALSE,
labels = NULL
)
calc_cro_sum(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL
)
calc_cro_median(
cro_fun

```r
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL
)
calc_cro_pearson(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL
)
calc_cro_spearman(
data,
cell_vars,
col_vars = total(),
row_vars = total(label = ""),
weight = NULL,
subgroup = NULL
)
combine_functions(..., method = c)
```

**Arguments**

- `cell_vars`: vector/data.frame/list. Variables on which summary function will be computed.
- `weight`: numeric vector. Optional cases weights. Cases with NA's, negative and zero weights are removed before calculations.
- `subgroup`: logical vector. You can specify subgroup on which table will be computed.
- `fun`: custom summary function. Generally it is recommended that `fun` will always return object of the same form. Rownames/vector names of `fun` result will appear in the row labels of the table and column names/names of list will appear in the column labels. To use weight you should have formal weight argument in `fun` and some logic for its processing inside. For `cro_fun_df` `fun` will receive `data.table` with all names converted to variable labels (if labels exists). So it is not recommended to rely on original variables names in your `fun`.
- `...`: further arguments for `fun` in `cro_fun/cro_fun_df` or functions for `combine_functions`. Ignored in `cro_fun/cro_fun_df` if `unsafe` is TRUE.
unsafe  logical/character If not FALSE than fun will be evaluated as is. It can lead to significant increase in the performance. But there are some limitations. For cro_fun it means that your function fun should return vector. If length of this vector is greater than one than you should provide with unsafe argument vector of unique labels for each element of this vector. There will be no attempts to automatically make labels for the results of fun. For cro_fun_df your function should return vector or list/data.frame (optionally with 'row_labels' element - statistic labels). If unsafe is TRUE or not logical then further arguments (...) for fun will be ignored.

weighted_valid_n  logical. Should we show weighted valid N in cro_mean_sd_n? By default it is FALSE.

labels  character vector of length 3. Labels for mean, standard deviation and valid N in cro_mean_sd_n.

data  data.frame in which context all other arguments will be evaluated (for calc_cro_*).

method  function which will combine results of multiple functions in combine_functions. It will be applied as in the expression do.call(method,list_of_functions_results). By default it is c.

Value

object of class 'etable'. Basically it's a data.frame but class is needed for custom methods.

See Also
tables, fre, cro.

Examples

data(mtcars)

```r
mtcars = apply_labels(mtcars,
                   mpg = "Miles/(US) gallon",
                   cyl = "Number of cylinders",
                   disp = "Displacement (cu.in.)",
                   hp = "Gross horsepower",
                   drat = "Rear axle ratio",
                   wt = "Weight (1000 lbs)",
                   qsec = "1/4 mile time",
                   vs = "Engine",
                   vs = c("V-engine" = 0,
                          "Straight engine" = 1),
                   am = "Transmission",
                   am = c("Automatic" = 0,
                          "Manual"=1),
                   gear = "Number of forward gears",
                   carb = "Number of carburetors"
)
```

# Simple example - there is special shortcut for it - 'cro_mean'
calculate(mtcars, cro_fun(list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = vs,
    fun = mean)
)

# the same result
calc_cro_fun(mtcars, list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = vs,
    fun = mean)

# The same example with 'subgroup'
calculate(mtcars, cro_fun(list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = vs,
    subgroup = vs == 0,
    fun = mean)

# 'combine_functions' usage
calculate(mtcars, cro_fun(list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = vs,
    fun = combine_functions(Mean = mean,
        'Std. dev.' = sd,
        'Valid N' = valid_n)
    ))

# 'combine_functions' usage - statistic labels in columns
calculate(mtcars, cro_fun(list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = vs,
    fun = combine_functions(Mean = mean,
        'Std. dev.' = sd,
        'Valid N' = valid_n,
        method = list)
    ))

# 'summary' function
calculate(mtcars, cro_fun(list(mpg, disp, hp, wt, qsec),
    col_vars = list(total(), am),
    row_vars = list(total(), vs),
    fun = summary)

# comparison 'cro_fun' and 'cro_fun_df'
calculate(mtcars, cro_fun(
    sheet(mpg, disp, hp, wt, qsec),
    col_vars = am,
    fun = mean)
)

# same result
calculate(mtcars, cro_fun_df(
    sheet(mpg, disp, hp, wt, qsec),
    col_vars = am,
    fun = colMeans
  )
)

# usage for 'cro_fun_df' which is not possible for 'cro_fun'
# linear regression by groups
calculate(mtcars, cro_fun_df(
    sheet(mpg, disp, hp, wt, qsec),
    col_vars = am,
    fun = function(x){
      frm = reformulate(".", response = names(x)[1])
      model = lm(frm, data = x)
      sheet(
        'Coef. estimate' = coef(model),
        confint(model)
      )
    }
  )
)

---

default_dataset  

Get or set reference to default dataset. Experimental feature.

Description

Use data.frame or data.frame name to set it as default. Use NULL as an argument to disable default dataset. If argument is missing then function will return reference to default dataset. Use ref to modify it. Also see .compute for usage patterns.

Usage

default_dataset(x)

Arguments

x  
data.frame or data.frame name which we want to make default for some operations.

Value

formula reference to default dataset or NULL

See Also

ref
do_repeat

Examples

data(iris)
default_iris = iris
default_dataset(default_iris) # set default dataset

.computer({
    new_col = 1
    Sepal.Length = Sepal.Length*2
})

# for comparison
iris$new_col = 1
iris$Sepal.Length = iris$Sepal.Length*2
identical(iris, default_iris) # should be TRUE

default_dataset(NULL) # disable default dataset

---
do_repeat

Repeats the same transformations on a specified set of variables/values

Description
Repeats the same transformations on a specified set of variables/values

Usage

do_repeat(data, ...)

.do_repeat(...)

as_is(...)

Arguments

data data.frame/list. If data is list then do_repeat will be applied to each element of the list.

... stand-in name(s) followed by equals sign and a vector/list of replacement variables or values. They can be numeric/characters or variables names. Names at the top-level can be unquoted (non-standard evaluation). Quoted characters also considered as variables names. To avoid this behavior use as_is function. For standard evaluation of parameters you can surround them by round brackets. Also you can use %to% operator and other criteria functions. Last argument should be expression in curly brackets which will be evaluated in the scope of data.frame data. See examples.
Details

There is a special constant `.N` which equals to number of cases in `data` for usage in expression inside `do_repeat`. Also there are a variables `.item_num` which is equal to the current iteration number and `.item_value` which is named list with current stand-in variables values.

Value

transformed data.frame data

See Also

compute, do_if

Examples

data(iris)
scaled_iris = do_repeat(iris,
i = Sepal.Length %to% Petal.Width,
{   i = scale(i)
    })
head(scaled_iris)

# several stand-in names and standard evaluation
old_names = qc(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width)
new_names = paste0("scaled_", old_names)
scaled_iris = do_repeat(iris,
    orig = ((old_names)),
scaled = ((new_names)),
    {    scaled = scale(orig)
    })
head(scaled_iris)

# numerics
new_df = data.frame(id = 1:20)
# note the automatic creation of the sequence of variables
new_df = do_repeat(new_df,
    item = i1 %to% i3,
    value = c(1, 2, 3),
    {    item = value
    })
head(new_df)

# the same result with internal variable `.item_num'
new_df = data.frame(id = 1:20)
nuew_df = do_repeat(new_df,
    item = i1 %to% i3,
    {    item = .item_num
    })
head(new_df)

# functions
set.seed(123)
new_df = data.frame(id = 1:20)
new_df = do_repeat(new_df,
item = c(i1, i2, i3),
fun = c("rnorm", "runif", "rexp"),
{
    item = fun(.N)
})
head(new_df)

---

**drop_empty_rows**

*Drop empty (with all NA's) rows/columns from data.frame/table*

**Description**

By default tables produced by functions `tables`, `cro`, `cro_fun` and `cro_fun_df` are created with all possible value labels. If values for this labels are absent in variable there are NA's in rows and columns. `drop_empty_rows`/`drop_empty_columns` are intended to remove these empty rows/columns. `drop_r` and `drop_c` are the same functions with shorter names. `drop_rc` drops rows and columns simultaneously.

**Usage**

```
drop_empty_rows(x, excluded_rows = NULL, excluded_columns = NULL)
drop_empty_columns(x, excluded_rows = NULL, excluded_columns = NULL)
drop_r(x, excluded_rows = NULL, excluded_columns = NULL)
drop_c(x, excluded_rows = NULL, excluded_columns = NULL)
drop_rc(x)
```

**Arguments**

- `x` data.frame/etable(result of `cro` and etc.)
- `excluded_rows` character/logical/numeric rows which won’t be dropped and in which NAs won’t be counted. If it is characters then they will be considered as pattern/vector of patterns. Patterns will be matched with Perl-style regular expression with values in the first column of `x` (see `grep`, `perl = TRUE` argument). Rows which have such patterns will be excluded. By default for class 'etable' pattern is "#" because "#" marks totals in the result of `cro`. 

- `excluded_columns`
excluded_columns

logical/numeric/characters columns which won’t be dropped and in which NAs won’t be counted. By default for class ‘etable’ it is first column - column with labels in table.

Value

data.frame with removed rows/columns

Examples

data(mtcars)
mtcars = apply_labels(mtcars, vs = “Engine”,
vs = num_lab("0 V-engine
1 Straight engine
9 Other
"),
am = "Transmission",
am = num_lab("0 Automatic
1 Manual
9 Other
")
)
with_empty = calculate(mtcars, cro(am, vs))

drop_empty_rows(with_empty)
drop_empty_columns(with_empty)
drop_rc(with_empty)

experimental

Experimental functions for operations with default dataset

Description

Workflow for these functions is rather simple. You should set up default data.frame with default_dataset and then operate with it without any reference to your data.frame. There are two kinds of operations. The first kind modify default dataset, the second kind will be evaluated in the context of the default dataset but doesn’t modify it. It is not recommended to use one of these functions in the scope of another of these functions. By now their performance is not so high, especially .do_if/.modify_if can be very slow.

Usage

.compute(expr)
.do_if(cond, expr)
.modify_if(cond, expr)
.modify(expr)
.calculate(expr, use_labels = FALSE)
.calc(expr, use_labels = FALSE)
.val_lab(...) 
.var_lab(...) 
.recode(x, ...)
.fre(...)
.cro(...)
.cro_cases(...)
.cro_cpct(...)
.cro_rpct(...)
.cro_tpct(...)
.cro_mean(...)
.cro_sum(...)
.cro_median(...)
.cro_mean_sd_n(...)
.cro_fun(...)
.cro_fun_df(...)

Arguments

expr set of expressions in curly brackets which will be evaluated in the context of default dataset
cond logical vector/expression
use_labels logical. Experimental feature. If it equals to TRUE then we will try to replace variable names with labels. Many base R functions which show variable names will show labels.
... further arguments

x vector/data.frame - variable names in the scope of default dataset

Details

Functions which modify default dataset:

- `modify` Add and modify variables inside default data.frame. See modify.
- `compute` Shortcut for `modify`. Name is inspired by SPSS COMPUTE operator. See modify.
- `modify_if` Add and modify variables inside subset of default data.frame. See modify_if.
- `do_if` Shortcut for `modify_if`. Name is inspired by SPSS DO IF operator. See modify_if.
- `where` Leave subset of default data.frame which meet condition. See where, subset.
- `recode` Change, rearrange or consolidate the values of an existing variable inside default data.frame. See recode.

Other functions:

- `var_lab` Return variable label from default dataset. See var_lab.
- `val_lab` Return value labels from default dataset. See val_lab.
- `fre` Simple frequencies of variable in the default data.frame. See fre.
- `cro` Simple crosstabulations of variable in the default data.frame. See cro.
- `calculate` Evaluate arbitrary expression in the context of data.frame. See calculate.

Examples

data(mtcars)
default_dataset(mtcars) # set mtcars as default dataset

# calculate new variables
.compute({
mpg_by_am = ave(mpg, am, FUN = mean)
hi_low_mpg = ifs(mpg<mean(mpg) ~ 0, TRUE ~ 1)
})

# set labels
.apply_labels(
mpg = "Miles/(US) gallon",
cyl = "Number of cylinders",
disp = "Displacement (cu.in.)",
hp = "Gross horsepower",
mpg_by_am = "Average mpg for transmission type",
hi_low_mpg = "Miles per gallon",
hi_low_mpg = num_lab("0 Low


expss: Tables with Labels and Some Useful Functions from Spreadsheets and SPSS Statistics

1 High

vs = "Engine",
vs = num_lab("0 V-engine
1 Straight engine
"),

am = "Transmission",
am = num_lab("0 Automatic
1 Manual
")

)  
# calculate frequencies  
.fre(hi_low_mpg)  
cro(cyl, hi_low_mpg)  
cro_mean(data.frame(mpg, disp, hp), vs)

# disable default dataset  
default_dataset(NULL)

# Example of .recode  
data(iris)  
default_dataset(iris) # set iris as default dataset  
.recode(Sepal.Length, lo %thru% median(Sepal.Length) ~ "small", other ~ "large")  
.fre(Sepal.Length)

# example of .do_if  
.do_if(Species == "setosa",{
Petal.Length = NA
Petal.Width = NA
})

cro_mean(data.frame(Petal.Length, Petal.Width), Species)

# disable default dataset  
default_dataset(NULL)
Description

'expss' package implements some popular functions from spreadsheets and SPSS Statistics software. Implementations are not complete copies of their originals. I try to make them consistent with other R functions. See examples in the vignette and in the help.

Excel

- IF ifelse
- AVERAGE mean_row
- SUM sum_row
- MIN min_row
- MAX max_row
- VLOOKUP vlookup
- COUNTIF count_if
- AVERAGEIF mean_row_if
- SUMIF sum_row_if
- MINIF min_row_if
- MAXIF max_row_if
- IFS ifs
- IFNA if_na
- MATCH match_row
- INDEX index_row
- PIVOT TABLES tables, cro_fun, cro

SPSS

- COMPUTE compute
- RECODE recode
- COUNT count_row_if
- DO IF do_if
- DO REPEAT do_repeat
- VARIABLE LABELS var_lab
- VALUE LABELS val_lab
- ANY any_in_row
- FREQUENCIES fre
- CROSSTABS cro
- CUSTOM TABLES tables
Description

All options can be set with `options(option.name = option.value)` or with special functions (see below). You can get value of option with `getOption("option.name")`.

- **expss.digits** Number of digits after decimal separator which will be shown for tables. This parameter is supported in the `as.datatable_widget`, `htmlTable.table` and `print` methods. `NULL` is default and means one digit. `NA` means no rounding. There is a convenience function for this option: `expss_digits`.

- **expss.enable_value_labels_support** By default, all labelled variables will use labels as labels for factor levels when `factor` is called. So, any function which calls `factor/as.factor` will use value labels. In details this option changes behavior of two methods for class `labelled` - `as.character` and `unique` - on which `factor` depends entirely. If you have compatibility problems set this option to zero: `options(expss.enable_value_labels_support = 0)`. Additionally there is an option for extreme value labels support: `options(expss.enable_value_labels_support = 2)`. With this value `factor/as.factor` will take into account empty levels. See example. It is recommended to turn off this option immediately after usage because `unique.labelled` will give weird result. Labels without values will be added to unique values. There are shortcuts for these options: `expss_enable_value_labels_support()`, `expss_enable_value_labels_support_extreme()`, and `expss_disable_value_labels_support()`.

- **expss.output** By default tables are printed in the console. You can change this behavior by setting this option. There are five possible values: 'rnotebook', 'viewer', 'commented', 'raw' or 'huxtable'. First option is useful when you run your code in the R Notebook - output will be rendered to nice HTML. The second option will render tables to RStudio viewer. `knitr` is supported automatically via `knit_print` method. 'commented' prints default output to the console with comment symbol (#) at the beginning of the each line. With comment symbol you can easily copy and paste your output into the script. Option `raw` disables any formatting and all tables are printed as data.frames. Option `huxtable` print output via the huxtable library. Shortcuts for options: `expss_output_default()`, `expss_output_raw()`, `expss_output_viewer()`, `expss_output_commented()`, `expss_output_rnotebook()` and `expss_output_huxtable()`.

- **expss_fix_encoding_on/expss_fix_encoding_off** If you experience problems with character encoding in RStudio Viewer/RNotebooks under Windows try `expss_fix_encoding_on()`. In some cases, it can help.

Usage

```r
expss_digits(digits = NULL)
get_expss_digits()
expss_enable_value_labels_support()
```
expss_enable_value_labels_support_extreme()
expss_disable_value_labels_support()
expss_output_default()
expss_output_commented()
expss_output_raw()
expss_output_viewer()
expss_output_rnotebook()
expss_output_huxtable(...)
expss_fix_encoding_on()
expss_fix_encoding_off()

expss_fre_stat_lab(
  label = c("Count", "Valid percent", "Percent", "Responses, %", 
    "Cumulative responses, %")
)

Arguments

 digits integer. Number of digits after decimal point. NULL is default and means 1 digit. NA means no rounding.
... list of parameters for huxtable::set_default_properties. See set_default_properties.
label character vector of length 5. Default labels for fre.

Examples

# example of different levels of value labels support
my_scale = c(1, 2, 2, 2)
# note that we have label 'Hard to say' for which there are no values in 'my_scale'
val_lab(my_scale) = num_lab("  
  1 Yes  
  2 No  
  3 Hard to say  
")

# disable labels support
expss_disable_value_labels_support()
table(my_scale) # there is no labels in the result
unique(my_scale)
# default value labels support
expss_enable_value_labels_support()
# table with labels but there are no label "Hard to say"
fctr

Convert labelled variable to factor

Description

fctr converts variable to factor. It force labels usage as factor labels for labelled variables even if `expss.enable_value_labels_support` set to 0. For other types of variables base factor is called. Factor levels are constructed as values labels. If label doesn’t exist for particular value then this value remain as is - so there is no information lost. This levels look like as “Variable_label|Value label” if argument prepend set to TRUE.

Usage

fctr(x, ..., drop_unused_labels = FALSE, prepend_var_lab = TRUE)

Arguments

x a vector of data with labels.

... optional arguments for factor

drop_unused_labels logical. Should we drop unused value labels? Default is FALSE.

prepend_var_lab logical. Should we prepend variable label before value labels? Default is TRUE.

Value

an object of class factor. For details see base factor documentation.

See Also

values2labels, names2labels, val_lab, var_lab. Materials for base functions: factor, as.factor, ordered, as.ordered
Examples

```r
data(mtcars)

var_lab(mtcars$am) = "Transmission"
val_lab(mtcars$am) = c(automatic = 0, manual=1)

summary(lm(mpg ~ am, data = mtcars)) # no labels
summary(lm(mpg ~ fctr(am), data = mtcars)) # with labels
summary(lm(mpg ~ fctr(unvr(am)), data = mtcars)) # without variable label
```

**Description**

The `fre` function returns a data.frame with six columns: labels or values, counts, valid percent (excluding NA), percent (with NA), percent of responses (for single-column `x` it equals to valid percent) and cumulative percent of responses.

**Usage**

```r
fre(
  x,
  weight = NULL,
  drop_unused_labels = TRUE,
  prepend_var_lab = FALSE,
  stat_lab = getOption("expss.fre_stat_lab", c("Count", "Valid percent", "Percent", "Responses, %", "Cumulative responses, %"))
)
```

**Arguments**

- **x** vector/data.frame/list. data.frames are considered as multiple response variables. If `x` is list then vertically stacked frequencies for each element of list will be generated.
- **weight** numeric vector. Optional case weights. NA's and negative weights treated as zero weights.
- **drop_unused_labels** logical. Should we drop unused value labels? Default is TRUE.
- **prepend_var_lab** logical. Should we prepend variable label before value labels? By default we will add variable labels to value labels only if `x` or predictor is list (several variables).
- **stat_lab** character. Labels for the frequency columns.
Value

object of class 'etable'. Basically it's a data.frame but class is needed for custom methods.

Examples

data(mtcars)
mtcars = modify(mtcars,
    var_lab(vs) = "Engine"
    val_lab(vs) = c("V-engine" = 0,
    "Straight engine" = 1)
    var_lab(am) = "Transmission"
    val_lab(am) = c(automatic = 0,
    manual=1)
)

fre(mtcars$vs)

# stacked frequencies
fre(list(mtcars$vs, mtcars$am))

# multiple-choice variable
# brands – multiple response question
# Which brands do you use during last three months?
set.seed(123)
brands = data.frame(t(replicate(20,sample(c(1:5,NA),4,replace = FALSE))))
# score – evaluation of tested product
score = sample(-1:1,20,replace = TRUE)
var_lab(brands) = "Used brands"
val_lab(brands) = make_labels("1 Brand A
2 Brand B
3 Brand C
4 Brand D
5 Brand E
")

var_lab(score) = "Evaluation of tested brand"
val_lab(score) = make_labels("-1 Dislike it
0 So-so
1 Like it
")

fre(brands)

# stacked frequencies
fre(list(score, brands))
htmlTable.etable  Outputting HTML tables in RStudio viewer/R Notebooks

Description

This is method for rendering results of fre/cro/tables in Shiny/RMarkdown/Jupyter notebooks and etc. For detailed description of function and its arguments see htmlTable. You can pack your tables in the list and render them all simultaneously. See examples. You may be interested in expss_output_viewer() for automatical rendering tables in the RStudio viewer or expss_output_rnotebook() for rendering in the R notebooks. See expss.options. repr_html is method for rendering table in the Jupyter notebooks and knit_print is method for rendering table in the knitr HTML-documents. Jupyter notebooks and knitr documents are supported automatically but in the R notebooks it is needed to set output to notebook via expss_output_rnotebook().

Usage

```r
## S3 method for class 'etable'
htmlTable(
x,
header = NULL,
rnames = NULL,
rowlabel = NULL,
caption = NULL,
tfoot = NULL,
label = NULL,
rgroup = NULL,
n.rgroup = NULL,
cgroup = NULL,
n.cgroup = NULL,
tspanner = NULL,
n.tspanner = NULL,
total = NULL,
ctable = TRUE,
compatibility = getOption("htmlTableCompat", "LibreOffice"),
cspan.rgroup = "all",
escape.html = FALSE,
..., 
digits = get_expss_digits(),
row_groups = TRUE
)
## S3 method for class 'with_caption'
htmlTable(
x,
header = NULL,
rnames = NULL,
rowlabel = NULL,
```
caption = NULL,
tfoot = NULL,
label = NULL,
rgroup = NULL,
n.rgroup = NULL,
cgroup = NULL,
n.cgroup = NULL,
tspanner = NULL,
n.tspanner = NULL,
total = NULL,
ctable = TRUE,
compatibility = getOption("htmlTableCompat", "LibreOffice"),
cspan.rgroup = "all",
escape.html = FALSE,
...
digits = get_exprs_digits(),
row_groups = TRUE
)

## S3 method for class 'list'
htmlTable(
x,
  header = NULL,
rnames = NULL,
rowlabel = NULL,
caption = NULL,
tfoot = NULL,
label = NULL,
rgroup = NULL,
n.rgroup = NULL,
cgroup = NULL,
n.cgroup = NULL,
tspanner = NULL,
n.tspanner = NULL,
total = NULL,
ctable = TRUE,
compatibility = getOption("htmlTableCompat", "LibreOffice"),
cspan.rgroup = "all",
escape.html = FALSE,
...
digits = get_exprs_digits(),
row_groups = TRUE,
gap = "<br>"
)

knit_print.etable(x, ..., digits = get_exprs_digits(), escape.html = FALSE)

knit_print.with_caption(
htmlTable.etable

```r
x,
..., 
digits = get_expss_digits(),
escape.html = FALSE
)

repr_html.etable(obj, ..., digits = get_expss_digits(), escape.html = FALSE)

repr_html.with_caption(
  obj,
  ..., 
  digits = get_expss_digits(),
  escape.html = FALSE
)

repr_text.etable(obj, ..., digits = get_expss_digits())

repr_text.with_caption(obj, ..., digits = get_expss_digits())

Arguments

- `x`: a data object of class 'etable' - result of fre/cro and etc.
- `header`: Ignored.
- `rnames`: Ignored.
- `rowlabel`: Ignored.
- `caption`: See manual for `htmlTable`.
- `tfoot`: See manual for `htmlTable`.
- `label`: See manual for `htmlTable`.
- `rgroup`: Ignored.
- `n.rgroup`: Ignored.
- `cgroup`: Ignored.
- `n.cgroup`: Ignored.
- `tspanner`: See manual for `htmlTable`.
- `n.tspanner`: See manual for `htmlTable`.
- `total`: See manual for `htmlTable`.
- `ctable`: See manual for `htmlTable`.
- `compatibility`: See manual for `htmlTable`.
- `cspan.rgroup`: See manual for `htmlTable`.
- `escape.html`: logical: should HTML characters be escaped? Defaults to FALSE.
- `...`: further parameters for `htmlTable`.
- `digits`: integer By default, all numeric columns are rounded to one digit after decimal separator. Also you can set this argument by setting option 'expss.digits' - for example, `expss_digits(2)`. If it is NA than all numeric columns remain un-rounded.
htmlTable.etable

row_groups: logical. Should we create row groups? TRUE by default.

gap: character. Separator between tables if we output list of tables. By default it is line break '<br>'.

obj: a data object of class 'etable' - result of fre/cro and etc.

Value

Returns a string of class htmlTable

Examples

```r
## Not run:
data(mtcars)
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (1000 lbs)",
  qsec = "1/4 mile time",
  vs = "Engine",
  am = "Transmission",
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

expss_output_viewer()

mtcars %>%
  tab_cols(total(), am %nest% vs) %>%
  tab_cells(mpg, hp) %>%
  tab_stat_mean() %>%
  tab_cells(cyl) %>%
  tab_stat_cpct() %>%
  tab_pivot() %>%
  set_caption("Table 1. Some variables from mtcars dataset.")

# several tables in a list
mtcars %>%
calc(list(
  cro_cpct(list(am, vs, cyl), list(total(), am)) %>%
  set_caption("Table 1. Percent."),
  cro_mean_sd_n(list(mpg, hp, qsec), list(total(), am)) %>%
  set_caption("Table 2. Means."))
)%>%
htmlTable()

expss_output_default()
## End(Not run)

### if_na

Replace values with NA and vice-versa

**Description**

- `if_na` replaces NA values in vector/data.frame/matrix/list with supplied value. For single value argument label can be provided with `label` argument. If replacement value is vector then `if_na` uses for replacement values from appropriate positions. An opposite operation is `na_if`.
- `na_if` replaces values with NA in vector/data.frame/matrix/list. Another alias for this is `mis_val`.
- `valid` returns logical vector which indicate the presence of at least one not-NA value in row. For vector or single column data.frame result is the same as with `complete_cases`. There is a special case for data.frame of class `dichotomy`. In this case result indicate the presence of at least one 1 in a row.

**Usage**

```r
if_na(x, value, label = NULL)
if_na(x, label = NULL) <- value
x %if_na% value
na_if(x, value, with_labels = FALSE)
na_if(x, with_labels = FALSE) <- value
x %na_if% value
mis_val(x, value, with_labels = FALSE)
mis_val(x, with_labels = FALSE) <- value
valid(x)
```

**Arguments**

- `x` vector/matrix/data.frame/list
- `value` single value, vector of the same length as number of rows in `x`, or function (`criteria`) for `na_if`. See `recode` for details.
- `label` a character of length 1. Label for value which replace NA.
- `with_labels` logical. FALSE by default. Should we also remove labels of values which we recode to NA?
Format

An object of class character of length 1.

Value

object of the same form and class as x. valid returns logical vector.

Examples

# simple case
a = c(NA, 2, 3, 4, NA)
if_na(a, 99)

# the same result
a %if_na% 99

# with label
a = c(NA, 2, 3, 4, NA)
if_na(a, 99, label = "Hard to say")

# in-place replacement. The same result:
if_na(a, label = "Hard to say") = 99
a # c(99, 2, 3, 4, 99)

# replacement with values from other variable
a = c(NA, 2, 3, 4, NA)
b = 1:5
if_na(a, b)

# replacement with group means
# make data.frame
set.seed(123)
group = sample(1:3, 30, replace = TRUE)
param = runif(30)
param[sample(30, 10)] = NA # place 10 NA's
df = data.frame(group, param)

# replace NA's with group means
if_na(df$param) = window_fun(df$param, df$group, mean_col)
df

# replace NA's with group means

### na_if examples ###

a = c(1:5, 99)
# 99 to NA
na_if(a, 99) # c(1:5, NA)
a %na_if% 99 # same result

# values which greater than 4 to NA
na_if(a, gt(4))  # c(1:4, NA, NA)

# alias 'mis_val', with_labels = TRUE
a = c(1, 1, 2, 2, 99)
val_lab(a) = c(Yes = 1, No = 2, "Hard to say" = 99)
mis_val(a, 99, with_labels = TRUE)

set.seed(123)
dfs = data.frame(
    a = c("bad value", "bad value", "good value", "good value", "good value"),
    b = runif(5)
)

# rows with 'bad value' will be filled with NA
# logical argument and recycling by columns
na_if(dfs, dfs$a=="bad value")

a = rnorm(50)
# values greater than 1 or less than -1 will be set to NA
# special functions usage
na_if(a, lt(-1) | gt(1))

# values inside [-1, 1] to NA
na_if(a, -1 %thru% 1)

info

Provides variables description for dataset

Description

info returns data.frame with variables description and some summary statistics. Resulting data.frame mainly intended to keep in front of eyes in RStudio viewer or to be saved as csv to view in the spreadsheet software as reference about working dataset.

Usage

info(x, stats = TRUE, frequencies = TRUE, max_levels = 10)

Arguments

x 
vector/factor/list/data.frame.

stats 
Logical. Should we calculate summary for each variable?

frequencies 
Logical. Should we calculate frequencies for each variable? This calculation can take significant amount of time for large datasets.

max_levels 
Numeric. Maximum levels for using in frequency calculations. Levels above this value will convert to 'Other values'.
Value

data.frame with following columns: Name, Class, Length, NotNA, NA, Distincts, Label, ValueLabels, Min., 1st Qu., Median, Mean, 3rd Qu., Max., Frequency.

Examples

data(mtcars)
var_lab(mtcars$am) = "Transmission"
val_lab(mtcars$am) = c("Automatic"=0, "Manual"=1)
info(mtcars, max_levels = 5)

keep
Keep or drop elements by name/criteria in data.frame/matrix

Description
keep selects variables/elements from data.frame by their names or by criteria (see criteria). except drops variables/elements from data.frame by their names or by criteria. Names at the top-level can be unquoted (non-standard evaluation). For standard evaluation of parameters you can surround them by round brackets. See examples. Methods for list will apply keep/except to each element of the list separately. .keep/.except are versions which works with default_dataset.

Usage
keep(data, ...)
.keep(...)
except(data, ...)
.except(...)

Arguments
data data.frame/matrix/list
...
column names of type character/numeric or criteria/logical functions

Value
object of the same type as data
**Examples**

```r
data(iris)
keep(iris, Sepal.Length, Sepal.Width)
except(iris, Species)

keep(iris, Species, other()) # move 'Species' to the first position
keep(iris, to("Petal.Width")) # keep all columns up to 'Species'

except(iris, perl("^Petal")) # remove columns which names start with 'Petal'
except(iris, 5) # remove fifth column

data(mtcars)
keep(mtcars, from("mpg") & to("qsec")) # keep columns from 'mpg' to 'qsec'
keep(mtcars, mpg %to% qsec) # the same result

# standard and non-standard evaluation
many_vars = c("am", "vs", "cyl")
## Not run:
keep(mtcars, many_vars) # error - names not found: 'many_vars'

## End(Not run)
keep(mtcars, (many_vars)) # ok

# character expansion
dfs = data.frame(
a = 10 %r% 5,
b_1 = 11 %r% 5,
b_2 = 12 %r% 5,
b_3 = 12 %r% 5,
b_4 = 14 %r% 5,
b_5 = 15 %r% 5
)
i = 1:5
keep(dfs, b_1 %to% b_5)
keep(dfs, text_expand("b_{i}")) # the same result
```

---

**match_row**

Match finds value in rows or columns/index returns value by index from rows or columns

---

**Description**

match finds value in rows or columns. index returns value by index from row or column. One can use functions as criteria for match. In this case position of first value on which function equals to TRUE will be returned. For convenience there are special predefined functions - see criteria. If value is not found then NA will be returned.
Usage

match_row(criterion, ...)
match_col(criterion, ...)

index_row(index, ...)

index_col(index, ...)

value_row_if(criterion, ...)

value_col_if(criterion, ...)

Arguments

criterion Vector of values to be matched, or function.

... data. Vectors, matrices, data.frames, lists. Shorter arguments will be recycled.

index vector of positions in rows/columns from which values should be returned.

Value

vector with length equals to number of rows for *_row and equals to number of columns for *_col.

Examples

# toy data
v1 = 1:3
v2 = 2:4
v3 = 7:5

# positions of 1,3,5 in rows
match_row(c(1, 3, 5), v1, v2, v3) # 1:3
# positions of 1,3,5 in columns
match_col(1, v1, v2, v3) # c(v1 = 1, v2 = NA, v3 = NA)

# position of first value greater than 2
ix = match_row(gt(2), v1, v2, v3)
ix # c(3,2,1)
# return values by result of previous 'match_row'
index_row(ix, v1, v2, v3) # c(7,3,3)

# the same actions with data.frame
dfs = data.frame(v1, v2, v3)

# positions of 1,3,5 in rows
match_row(c(1, 3, 5), dfs) # 1:3
# positions of 1,3,5 in columns
match_col(1, dfs) # c(v1 = 1, v2 = NA, v3 = NA)

# position of first value greater than 2
merge.etable

Merge two tables/data.frames

Description

%merge% is infix shortcut for base merge with all.x = TRUE and all.y = FALSE (left join). There is also special method for combining results of cro_* and fre. For them all = TRUE (full join). It allows make complex tables from simple ones. See examples. Strange result is possible if one or two arguments have duplicates in first column (column with labels).

Usage

x %merge% y

Arguments

x data.frame or results of fre/cro_*/table_*
y data.frame or results of fre/cro_*/table_*

Value
data.frame

See Also

fre, cro, cro, cro_fun, merge

Examples

data(mtcars)
# apply labels
mtcars = apply_labels(mtcars,
    mpg = "Miles/(US) gallon",
cyl = "Number of cylinders",
disp = "Displacement (cu.in.)",
hp = "Gross horsepower",
drat = "Rear axle ratio",
wt = "Weight (lb/1000)",
qsec = "1/4 mile time",
vs = "V/S",
vs = c("V-engine" = 0, "Straight engine" = 1),
am = "Transmission (0 = automatic, 1 = manual)",
am = c(automatic = 0, manual = 1),
gear = "Number of forward gears",

ix = match_row(gt(2), dfs)
ix # c(3,2,1)
# return values by result of previous 'match_row'
index_row(ix, dfs) # c(7,3,3)
carb = "Number of carburetors"

# table by 'am'
tab1 = calculate(mtcars, cro_cpct(gear, am))
# table with percents
tab2 = calculate(mtcars, cro_cpct(gear, vs))

# combine tables
tabl %merge% tab2

# complex tables
# table with counts
counts = calculate(mtcars, cro(list(vs, am, gear, carb), list("Count")))
# table with percents
percents = calculate(mtcars, cro_cpct(list(vs, am, gear, carb), list("Column, %")))

# combine tables
counts %merge% percents

---

mrset

Create multiple response set/multiple dichotomy set from variables

Description

These functions are intended for usage with tables - `tables`, `cro`, `cro_fun`. Result of `mrset` is considered as multiple-response set with category encoding and result of `mdset` is considered as multiple response set with dichotomy (dummy) encoding e.g. with 0 or 1 in each column. Each column in the dichotomy is indicator of absence or presence of particular feature. Both functions don’t convert its arguments to anything - it is supposed that arguments already have appropriate encoding. For conversation see `as.dichotomy` or `as.category`.

- `mrset_f` and `mdset_f` select variables by fixed pattern. Fixed pattern can be unquoted. For details see `.f`.
- `mrset_p` and `mdset_p` select variables for multiple-responses by perl-style regular expression. For details see `.p`.
- `mrset_t` and `mdset_t` select variables by expanding text arguments. For details see `.t` and `text_expand`.

Usage

```r
mrset(..., label = NULL)
mdset(..., label = NULL)
mrset_f(..., label = NULL)
mrset_f(..., label = NULL)
```
mrset_p(..., label = NULL)
mdset_p(..., label = NULL)
mrset_t(..., label = NULL)
mdset_t(..., label = NULL)

Arguments

... variables
label character optional label for multiple response set

Value

data.frame of class category/dichotomy

See Also

as.dichotomy, as.category

Examples

data(product_test)
calc_cro_cpct(product_test, mrset(a1_1 %to% a1_6))
# same result
calc_cro_cpct(product_test, mrset_f(a1_))
# same result
calc_cro_cpct(product_test, mrset_p("a1_"))
# same result
calc_cro_cpct(product_test, mrset_t("a1_{1:6}"))

names2labels Replace data.frame/list names with corresponding variables labels.

Description

names2labels replaces data.frame/list names with corresponding variables labels. If there are no labels for some variables their names remain unchanged. n2l is just shortcut for names2labels.

Usage

names2labels(x, exclude = NULL, keep_names = FALSE)
n2l(x, exclude = NULL, keep_names = FALSE)
Arguments

- **x**: data.frame/list.
- **exclude**: logical/integer/character columns which names should be left unchanged. Only applicable to list/data.frame.
- **keep_names**: logical. If TRUE original column names will be kept with labels. Only applicable to list/data.frame.

Value

Object of the same type as x but with variable labels instead of names.

See Also

values2labels, val_lab, var_lab

Examples

```r
data(mtcars)
mtcars = modify(mtcars, {
  var_lab(mpg) = "Miles/(US) gallon"
  var_lab(cyl) = "Number of cylinders"
  var_lab(disp) = "Displacement (cu.in.)"
  var_lab(hp) = "Gross horsepower"
  var_lab(drat) = "Rear axle ratio"
  var_lab(wt) = "Weight (lb/1000)"
  var_lab(qsec) = "1/4 mile time"
  var_lab(vs) = "V/S"
  var_lab(am) = "Transmission (0 = automatic, 1 = manual)"
  var_lab(gear) = "Number of forward gears"
  var_lab(carb) = "Number of carburetors"
})
# without original names
# note: we exclude dependent variable 'mpg' from conversion to use its short name in formula
summary(lm(mpg ~ ., data = names2labels(mtcars, exclude = "mpg")))
# with names
summary(lm(mpg ~ ., data = names2labels(mtcars, exclude = "mpg", keep_names = TRUE)))
```

Description

Function is added to workaround strange bug with data.table (issue #10).

Usage

name_dots(...)
Arguments

... arguments

Value

list

---

`nest`  *Compute nested variable(s) from several variables*

Description

`nest` mainly intended for usage with table functions such as `cro`. See examples. `%nest%` is infix version of this function. You can apply `nest` on multiple-response variables/list of variables and data.frames.

Usage

`nest(...)`

`x %nest% y`

Arguments

... vectors/data.frames/lists
`x` vector/data.frame/list
`y` vector/data.frame/list

Value

vector/data.frame/list

See Also

See also `interaction`

Examples

data(mtcars)

`mtcars = apply_labels(mtcars,
    cyl = "Number of cylinders",
    vs = "Engine",
    vs = num_lab("0 V-engine
    1 Straight engine
    ")",
    am = "Transmission",
    ...)`
```r
am = num_lab("$
0 Automatic
1 Manual
",
)
carb = "Number of carburetors"

calc(mtcars, cro(cyl, am %nest% vs))
# list of variables
calc(mtcars, cro(cyl, am %nest% list(vs, cyl)))
# list of variables - multiple banners/multiple nesting
calc(mtcars, cro(cyl, list(total(), list(am, vs) %nest% cyl)))
# three variables
calc(mtcars, cro(am %nest% vs %nest% carb, cyl))
# the same with usual version
calc(mtcars, cro(cyl, nest(am, vs)))
# three variables
calc(mtcars, cro(nest(am, vs, carb), cyl))
```

---

**net**  
*Add subtotal to a set of categories*

**Description**

'Subtotal' adds subtotal to set of categories, 'net' replaces categories with their net value. If you provide named arguments then name will be used as label for subtotal. In other case labels will be automatically generated taking into account arguments 'new_label' and 'prefix'. Note that if you provide overlapping categories then net and subtotals will also be overlapping. 'subtotal' and 'net' are intended for usage with cro and friends. 'tab_subtotal_*' and 'tab_net_*' are intended for usage with custom tables - see tables. There are auxiliary functions 'hide' and 'unhide'. 'hide' is used with 'subtotal' when you need to leave only subtotal for some specific items. And 'unhide' is used with 'net' when you want to show items for some nets. See examples.

**Usage**

```r
net(
  x,
  ...
)
position = c("below", "above", "top", "bottom"),
prefix = "TOTAL ",
new_label = c("all", "range", "first", "last"),
add = FALSE
)```
subtotal(
  x,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last"),
  add = TRUE
)

tab_net_cells(
  data,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)

tab_net_cols(
  data,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)

tab_net_rows(
  data,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)

tab_subtotal_cells(
  data,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)

tab_subtotal_cols(
  data,
  ...
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)
net

```r
tab_subtotal_rows(
  data,
  ...,
  position = c("below", "above", "top", "bottom"),
  prefix = "TOTAL ",
  new_label = c("all", "range", "first", "last")
)

hide(category)
unhide(category)
```

**Arguments**

- `x` variable, list, data.frame or multiple response set
- `...` list of categories for grouping. It can be numeric vectors (for example, 1:2), ranges (for example, 4 greater(5)). If an argument is named then this name will be used as label for subtotal.
- `position` position of the subtotal or net relative to original categories. "below" by default. One of the "below", "above", "top", "bottom". "top" and "bottom" place nets and subtotals above or below all other categories. For nets "below" and "above" have no difference because original categories are removed.
- `prefix` character, "TOTAL " by default. It is a prefix to automatically created labels for nets and subtotals.
- `new_label` how we will combine original values for automatically generated subtotal labels. Possible values are "all", "range", "first", "last". "all" collapse all labels, "range" take only first and last label.
- `add` logical. Should we add subtotal to categories or replace categories with a net?
- `data` intermediate table. See `tables`.
- `category` category (numeric vectors, ranges, criteria) which you want to 'hide' or 'unhide'.

**Value**

multiple response set or list of the multiple response sets

**Examples**

```r
ol = c(1:7, 99)
var_lab(ol) = "Liking"
val_lab(ol) = num_lab("1 Disgusting
2 Very Poor
3 Poor
4 So-so
5 Good
6 Very good
```

7 Excellent
99 Hard to say
"

cro(subtotal(ol, BOTTOM = 1:3, TOP = 6:7, position = "top"))

# example with hide
cro(subtotal(ol, TOP1 = hide(7), TOP2 = hide(6:7), TOP3 = 5:7, BOTTOM = 1:3, position = "top"))

# autolabelling
cro(subtotal(ol, 1:3, 6:7))

# replace original codes and another way of autolabelling
cro(net(ol, 1:3, 6:7, new_label = "range", prefix = "NET "))

# unhide
cro(net(ol, 1:3, unhide(6:7), new_label = "range", prefix = "NET "))

# character variable and criteria usage
items = c("apple","banana","potato","orange","onion","tomato","pineapple")
cro(
    subtotal(items,
    "TOTAL FRUITS" = like("*ap*") | like("*an*"),
    "TOTAL VEGETABLES" = like("*to*") | like("*on*"),
    position = "bottom"
)
)


# 'tab_net_*' usage

data(mtcars)

mtcars = apply_labels(mtcars,
    mpg = "Miles/(US) gallon",
    am = "Transmission",
    am = c("Automatic" = 0,
        "Manual"=1),
    gear = "Number of forward gears",
    gear = c(
        One = 1,
        Two = 2,
        Three = 3,
        Four = 4,
        Five = 5
    )
)

mtcars %>%
    tab_cells(mpg) %>%
    tab_net_cells("Low mpg" = less(mean(mpg)), "High mpg" = greater_or_equal(mean(mpg))) %>%
    tab_cols(total(), am) %>%
    tab_stat_cases() %>%
    tab_pivot()

mtcars %>%
    tab_cells(mpg) %>%
    tab_rows(gear) %>%
    tab_subtotal_rows(1:2, 3:4, "5 and more" = greater(4)) %>%
    tab_stat_mean() %>%
    tab_pivot()
prepend_values

Prepend values/variable names to value/variable labels

Description
These functions add values/variable names as prefixes to value/variable labels. Functions which start with `tab_` intended for usage inside table creation sequences. See examples and tables. It is recommended to use `tab_prepend_*` at the start of sequence of tables creation. If you use it in the middle of the sequence then previous statements will not be affected.

Usage
```r
prepend_values(x)
prepend_names(x)
prepend_all(x)
tab_prepend_values(data)
tab_prepend_names(data)
tab_prepend_all(data)
```

Arguments
- `x` vector/data.frame. `prepend_names` can be applied only to data.frames.
- `data` data.frame/intermediate result of tables construction. See tables.

Value
original object with prepended names/values to labels

Examples
```r
data(mtcars)
mtcars = apply_labels(mtcars,
         mpg = "Miles/(US) gallon",
         cyl = "Number of cylinders",
         disp = "Displacement (cu.in.)",
         hp = "Gross horsepower",
         drat = "Rear axle ratio",
         wt = "Weight (lb/1000)",
         qsec = "1/4 mile time",
         vs = "Engine",
         vs = c("V-engine" = 0,
                "Straight engine" = 1),
         am = "Transmission",
)```
product_test

Data from product test of chocolate confectionary

Description

It is truncated dataset with data from product test of two samples of chocolate sweets. 150 respondents tested two kinds of sweets (codenames: VSX123 and SDF546). Sample was divided into two groups (cells) of 75 respondents in each group. In cell 1 product VSX123 was presented first and then SDF546. In cell 2 sweets were presented in reversed order. Questions about respondent impressions about first product are in the block A (and about second tested product in the block B). At the end of the questionnaire there is a question about preferences between sweets.
Usage

`product_test`

Format

A data frame with 150 rows and 18 variables:

- **id**  Respondent Id.
- **cell**  First tested product (cell number).
- **s2a**  Age.
- **a1_1**  What did you like in these sweets? Multiple response. First tested product.
- **a1_2**  (continue) What did you like in these sweets? Multiple response. First tested product.
- **a1_3**  (continue) What did you like in these sweets? Multiple response. First tested product.
- **a1_4**  (continue) What did you like in these sweets? Multiple response. First tested product.
- **a1_5**  (continue) What did you like in these sweets? Multiple response. First tested product.
- **a1_6**  (continue) What did you like in these sweets? Multiple response. First tested product.
- **a22**  Overall liking. First tested product.
- **b1_1**  What did you like in these sweets? Multiple response. Second tested product.
- **b1_2**  (continue) What did you like in these sweets? Multiple response. Second tested product.
- **b1_3**  (continue) What did you like in these sweets? Multiple response. Second tested product.
- **b1_4**  (continue) What did you like in these sweets? Multiple response. Second tested product.
- **b1_5**  (continue) What did you like in these sweets? Multiple response. Second tested product.
- **b1_6**  (continue) What did you like in these sweets? Multiple response. Second tested product.
- **b22**  Overall liking. Second tested product.
- **c1**  Preferences.

---

**prop**  
*Compute proportions from numeric vector/matrix/data.frame*

Description

`prop` returns proportion to sum of entire `x`. `prop_col` returns proportion to sum of each column of `x`. `prop_row` returns proportion to sum of each row of `x`. Non-numeric columns in the data.frame are ignored. NA's are also ignored.

Usage

`prop(x)`

`prop_col(x)`

`prop_row(x)`
Arguments

x numeric vector/matrix/data.frame

Value

the same structure as x but with proportions of original values from sum of original values.

Examples

a = c(25, 25, NA)
prop(a)

# data.frame with non-numeric columns
fac = factor(c("a", "b", "c"))
char = c("a", "b", "c")
dat = as.POSIXct("2016-09-27")
a = sheet(fac, a = c(25, 25, NA), b = c(100, NA, 50), char, dat)

prop(a)
prop_row(a)
prop_col(a)

# the same as result as with 'prop.table'
tbl = table(state.division, state.region)

prop(tbl)
prop_row(tbl)
prop_col(tbl)

qc Create vector of characters from unquoted strings (variable names)

Description

• qc It is often needed to address variables in the data.frame in the such manner: dfs[ ,c("var1","var2","var3")].
  qc ("quoted c") is a shortcut for the such cases to reduce keystrokes. With qc you can write:
  dfs[ ,qc(var1,var2,var3)].
• qe returns list of expression. It is useful to create substitution list for ..$arg.
• text_expand is simple string interpolation function. It searches in its arguments expressions
  in curly brackets {expr}, evaluate them and substitute with the result of evaluation. See
  examples.

Usage

qc(...)
qe(...)
text_expand(..., delim = c("\{"", "\}"))
read_spss

Arguments

... characters in text_expand, unquoted names of variables in qc or unquoted expressions in qe.

delim character vector of length 2 - pair of opening and closing delimiters for the templating tags. By default it is curly brackets. Note that delim will be used in the perl-style regular expression so you need to escape special characters, e.g. use "\{" instead of \\{".

Value

Vector of characters

Examples

```r
## qc
qc(a, b, c)
identical(qc(a, b, c), c("a", "b", "c"))

mtcars[, qc(am, mpg, gear)]

## text_expand
i = 1:5
text_expand("q(i)")

i = 1:3
j = 1:3
textexpand("q1_{i}_{j}")

data(iris)
textexpand("'iris' has {nrow(iris)} rows.")

## qe
qe(mrset(a1 %to% a6), mrset(b1 %to% b6), mrset(c1 %to% c6))
```

read_spss  

Read an SPSS Data File

Description

read_spss reads data from a file stored in SPSS *.sav format. It returns data.frame and never converts string variables to factors. Also it prepares SPSS values/variables labels for working with val_lab/var_lab functions. User-missings values are ignored. read_spss is simple wrapper around read.spss function from package foreign.

Usage

```r
read_spss(file, reencode = TRUE)
read_spss_to_list(file, reencode = TRUE)
```
Arguments

- **file**: Character string: the name of the file or URL to read.
- **reencode**: logical: should character strings be re-encoded to the current locale. The default is TRUE. NA means to do so in a UTF-8 locale, only. Alternatively, a character string specifying an encoding to assume for the file.

Value

- `read_spss` returns data.frame.
- `read_spss_to_list` returns list of variables from SPSS files.

See Also

- `read.spss` in package `foreign`, `val_lab`, `var_lab`

Examples

```r
## Not run:
w = read_spss("project.123.sav") # to data.frame
list_w = read_spss_to_list("project.123.sav") # to list

## End(Not run)
```

---

**recode**

Change, rearrange or consolidate the values of an existing or new variable. Inspired by the RECODE command from SPSS.

Description

`recode` change, rearrange or consolidate the values of an existing variable based on conditions. Design of this function inspired by RECODE from SPSS. Sequence of recodings provided in the form of formulas. For example, 1:2 ~ 1 means that all 1’s and 2’s will be replaced with 1. Each value will be recoded only once. In the assignment form `recode(...) = ...` of this function values which don’t meet any condition remain unchanged. In case of the usual form `... = recode(...)` values which don’t meet any condition will be replaced with NA. One can use values or more sophisticated logical conditions and functions as a condition. There are several special functions for usage as criteria - for details see `criteria`. Simple common usage looks like: `recode(x, 1:2 ~ -1, 3 ~ 0, 1:2 ~ 1, 99 ~ NA)`. For more information, see details and examples. The `ifs` function checks whether one or more conditions are met and returns a value that corresponds to the first TRUE condition. `ifs` can take the place of multiple nested `ifelse` statements and is much easier to read with multiple conditions. `ifs` works in the same manner as `recode` - e.g. with formulas. But conditions should be only logical and it doesn’t operate on multicolumn objects.
**Usage**

```r
recode(
  x,
  ..., with_labels = FALSE,
  new_label = c("all", "range", "first", "last")
)
```

```r
rec(x, ..., with_labels = TRUE, new_label = c("all", "range", "first", "last"))
```

```r
recode(x, with_labels = FALSE, new_label = c("all", "range", "first", "last")) <- value
```

```r
rec(x, with_labels = TRUE, new_label = c("all", "range", "first", "last")) <- value
```

```r
if_s(...)
```

```r
lo
```

```r
hi
```

```r
copy(x)
```

```r
from_to(from, to)
```

```r
values %into% names
```

**Arguments**

- **x**: vector/matrix/data.frame/list
- **...**: sequence of formulas which describe recodings. They are used when from/to arguments are not provided.
- **with_labels**: logical. FALSE by default for 'recode' and TRUE for 'rec'. Should we also recode value labels with the same recodings as variable?
- **new_label**: one of "all", "range", "first", or "last". If we recode value labels (with_labels = TRUE) how will we combine labels for duplicated values? "all" will use all labels, "range" will use first and last labels. See examples.
- **value**: list with formulas which describe recodings in assignment form of function/to list if from/to notation is used.
- **from**: list of conditions for values which should be recoded (in the same format as LHS of formulas).
- **to**: list of values into which old values should be recoded (in the same format as RHS of formulas).
- **values**: object(-s) which will be assigned to names for %into% operation. %into% supports multivalue assignments. See examples.
- **names**: name(-s) which will be given to values expression. For %into%.
Format

An object of class numeric of length 1.

An object of class numeric of length 1.

Details

Input conditions - possible values for left-hand side (LHS) of formula or element of from list:

- vector/single value All values in x which equal to elements of the vector in LHS will be replaced with RHS.
- function Values for which function gives TRUE will be replaced with RHS. There are some special functions for the convenience - see criteria.
- single logical value TRUE It means all other unrecoded values (ELSE in SPSS RECODE). All other unrecoded values will be changed to RHS of the formula or appropriate element of to.

Output values - possible values for right-hand side (RHS) of formula or element of to list:

- value replace elements of x. This value will be recycled across rows and columns of x.
- vector values of this vector will replace values in the corresponding position in rows of x. Vector will be recycled across columns of x.
- function This function will be applied to values of x which satisfy recoding condition. There is a special auxiliary function copy which just returns its argument. So, in the recode it just copies old value (COPY in SPSS RECODE). See examples.

%into% tries to mimic SPSS 'INTO'. Values from left-hand side will be assigned to right-hand side. You can use %to% expression in the RHS of %into%. See examples. lo and hi are shortcuts for -Inf and Inf. They can be useful in expressions with %thru%, e.g. 1 %thru% hi.

Value

object of the same form as x with recoded values

Examples

# examples from SPSS manual
# RECODE V1 TO V3 (0=1) (1=0) (2, 3=-1) (9=9) (ELSE=SYSMIS)
v1 = c(0, 1, 2, 3, 9, 10)
recode(v1) = c(0 ~ 1, 1 ~ 0, 2:3 ~ -1, 9 ~ 9, TRUE ~ NA)
v1

# RECODE QVAR(1 THRU 5=1)(6 THRU 10=2)(11 THRU HI=3)(ELSE=0).
qvar = c(1:20, NA, NA)
recode(qvar, 1 %thru% 5 ~ 1, 6 %thru% 10 ~ 2, 11 %thru% hi ~ 3, TRUE ~ 0)
# the same result
recode(qvar, 1 %thru% 5 ~ 1, 6 %thru% 10 ~ 2, ge(11) ~ 3, TRUE ~ 0)

# RECODE STRNGVAR ('A', 'B', 'C'='A')('D', 'E', 'F'='B')(ELSE='').
strngvar = LETTERS
recode(strngvar, c('A', 'B', 'C') ~ 'A', c('D', 'E', 'F') ~ 'B', TRUE ~ '')
# recode in place. Note that we recode only first six letters
recode(stringvar) = c(c('A', 'B', 'C') ~ 'A', c('D', 'E', 'F') ~ 'B')
stringvar

# RECODE AGE (MISSING=9) (18 THRU HI=1) (0 THRU 18=0) INTO VOTER.
age = c(NA, 2:40, NA)
voter = recode(age, NA ~ 9, 18 %thru% hi ~ 1, 0 %thru% 18 ~ 0)
voter

# the same result with '%into%
recode(age, NA ~ 9, 18 %thru% hi ~ 1, 0 %thru% 18 ~ 0) %into% voter2
voter2

# recode with adding labels
voter = recode(age, "Refuse to answer" = NA ~ 9,
    "Vote" = 18 %thru% hi ~ 1,
    "Don't vote" = 0 %thru% 18 ~ 0)
voter

# recoding with labels
ol = c(1:7, 99)
var_lab(ol) = "Liking"
val_lab(ol) = num_lab("1 Disgusting
2 Very Poor
3 Poor
4 So-so
5 Good
6 Very good
7 Excellent
99 Hard to say")
recode(ol, 1:3 ~ 1, 5:7 ~ 7, TRUE ~ copy, with_labels = TRUE)

# 'rec' is a shortcut for recoding with labels. Same result:
rec(ol, 1:3 ~ 1, 5:7 ~ 7, TRUE ~ copy)

# another method of combining labels
recode(ol, 1:3 ~ 1, 5:7 ~ 7, TRUE ~ copy, with_labels = TRUE, new_label = "range")

# example with from/to notation
# RECODE QVAR(1 THRU 5=1)(6 THRU 10=2)(11 THRU HI=3)(ELSE=0).
list_from = list(1 %thru% 5, 6 %thru% 10, ge(11), TRUE)
list_to = list(1, 2, 3, 0)
recode(qvar, from_to(list_from, list_to))

list_from = list(NA, 18 %thru% hi, 0 %thru% 18)
list_to = list("Refuse to answer" = 9, "Vote" = 1, "Don't vote" = 0)
voter = recode(age, from_to(list_from, list_to))
voter

# 'ifs' examples
a = 1:5
b = 5:1
ifs(b>3 ~ 1) # c(1, 1, NA, NA, NA)
ifs(b>3 ~ 1, TRUE ~ 3) # c(1, 1, 3, 3, 3)
ifs(b>3 ~ 1, a>4 ~ 7, TRUE ~ 3)  # c(1, 1, 3, 3, 7)
ifs(b>3 ~ a, TRUE ~ 42)          # c(1, 2, 42, 42, 42)

# advanced usage

# multiple assignment with '%into%'
set.seed(123)
x1 = runif(30)
x2 = runif(30)
x3 = runif(30)
# note nessesary brackets around RHS of '%into%
recode(x1 %to% x3, gt(0.5) ~ 1, other ~ 0) %into% (x_rec_1 %to% x_rec_3)
fre(x_rec_1)
# the same operation with characters expansion
i = 1:3
recode(x1 %to% x3, gt(0.5) ~ 1, other ~ 0) %into% text_expand('x_rec2_[i]')
fre(x_rec2_1)

# factor recoding
a = factor(letters[1:4])
recode(a, "a" ~ "z", TRUE ~ copy) # we get factor

# example with function in RHS
data(iris)
new_iris = recode(iris, is.numeric ~ scale, other ~ copy)
str(new_iris)

set.seed(123)
a = rnorm(20)
# if a<(-0.5) we change it to absolute value of a (abs function)
recode(a, lt(-0.5) ~ abs, other ~ copy)
# the same example with logical criteria
recode(a, when(a<(-.5)) ~ abs, other ~ copy)

---

**ref**

**Auxiliary functions to pass arguments to function by reference**

**Description**

These two functions aimed to simplify build functions with side-effects (e.g. for modifying variables in place). Of course it is not the R way of doing things but sometimes it can save several keystrokes.

**Usage**

ref(x)

ref(x) <- value
Arguments

x Reference to variable, it is formula, ~var_name.
value Value that should be assigned to modified variable.

Details

To create reference to variable one can use formula: b = ~a. b is reference to a. So ref(b) returns value of a and ref(b) = new_val will modify a. If argument x of these functions is not formula then these functions have no effect e. g. ref(a) is identical to a and after ref(a) = value a is identical to value. It is not possible to use function as argument x in assignment form. For example, ref(some_function(x)) = some_value will rise error. Use y = some_function(x); ref(y) = some_value instead.

Value

ref returns value of referenced variable. ref<- modifies referenced variable.

Examples

# Simple example
a = 1:3
b = ~a # b is reference to 'a'
identical(ref(b), a) # TRUE

ref(b)[2] = 4 # here we modify 'a'
identical(a, c(1, 4, 3)) # TRUE

# usage inside function

# top 10 rows
head10 = function(x){
  ds = head(ref(x), 10)
  ref(x) = ds
  invisible(ds) # for usage without references
}
data(iris)
ref_to_iris = ~iris
head10(ref_to_iris) # side-effect
nrow(iris) # 10

# argument is not formula - no side-effect
data(mtcars)
mtcars10 = head10(mtcars)
nrow(mtcars10) # 10
nrow(mtcars) # 32
Description

To drop caption use `set_caption` with `caption = NULL`. Captions are supported by `htmlTable.etable`, `xl_write` and `datatable_widget` functions.

Usage

```r
set_caption(obj, caption)
get_caption(obj)
is.with_caption(obj)
```

Arguments

- `obj` object of class `etable` - result of `cro_cpct` and etc.
- `caption` character caption for the table.

Value

object of class `with_caption`.

Examples

```r
data(mtcars)
mtcars = apply_labels(mtcars,
                      vs = "Engine",
                      vs = num_lab(")
                      \ 0 V-engine
                      \ 1 Straight engine
                      "),
am = "Transmission",
am = num_lab(")
                      \ 0 Automatic
                      \ 1 Manual
                      
)
tbl_with_caption = calc_cro(mtcars, am, vs) %>%
                  set_caption("Table 1. Type of transmission.")

tbl_with_caption
```
**sheet**

Make data.frame without conversion to factors and without fixing names

---

**Description**

`sheet` and `as.sheet` are shortcuts to `data.frame` and `as.data.frame` with `stringsAsFactors = FALSE`, `check.names = FALSE`. `.sheet` is the same as above but works in the scope of default dataset.

**Usage**

```r
sheet(...)
as.sheet(x, ...)
.sheet(...)
```

**Arguments**

- `...` objects, possibly named
- `x` object to be coerced to data.frame

**Value**

data.frame/list

**See Also**

default_dataset, data.frame, as.data.frame

**Examples**

```r
# see the difference
df1 = data.frame(a = letters[1:3], "This is my long name" = 1:3)
df2 = sheet(a = letters[1:3], "This is my long name" = 1:3)

str(df1)
str(df2)

data(iris)
default_dataset(iris)
.sheet(Sepal.Width, Sepal.Length)
```
sort_asc  

Sort data.frames/matrices/vectors

Description

sort_asc sorts in ascending order and sort_desc sorts in descending order. .sort_asc/.sort_desc are versions for working with default_dataset.

Usage

sort_asc(data, ..., na.last = FALSE)

.sort_asc(..., na.last = FALSE)

sort_desc(data, ..., na.last = TRUE)

.sort_desc(..., na.last = TRUE)

Arguments

data  data.frame/matrix/vector

...  character/numeric or criteria/logical functions (see criteria). Column names/numbers for data.frame/matrix by which object will be sorted. Names at the top-level can be unquoted (non-standard evaluation). For standard evaluation of parameters you can surround them by round brackets. See examples. Ignored for vectors.

na.last  for controlling the treatment of NAs. If TRUE, missing values in the data are put last; if FALSE, they are put first; if NA, they are removed.

Value

sorted data

Examples

data(mtcars)
sort_asc(mtcars, mpg)
sort_asc(mtcars, cyl, mpg) # by two column

# same results with column nums
sort_asc(mtcars, 1)
sort_asc(mtcars, 2:1) # by two column
sort_asc(mtcars, 2, 1) # by two column

# call with parameter
sorting_columns = c("cyl", "mpg")
sort_asc(mtcars, (sorting_columns))
split_by

Splits data.frame into list of data.frames that can be analyzed separately.

Description

Splits data.frame into list of data.frames that can be analyzed separately. These data.frames are sets of cases that have the same values for the specified split variables. Any missing values in split variables are dropped together with the corresponding values of data. split_off works with lists of data.frames or objects that can be coerced to data.frame and assumed to have compatible structure. Resulting rows will be sorted in order of the split variables.

Usage

split_by(data, ..., drop = TRUE)

split_off(data, groups = NULL, rownames = NULL)

Arguments

data: data.frame for split_by/list for split_off
...
unquoted variables names (see keep) by which data will be split into list.
drop: should we drop combination of levels with zero observation? TRUE by default.
groups: character If it is not NULL then we add list names as variable to result of split_off with the name specified by groups. If it is TRUE then name will be .groups.
rownames: character If it is not NULL then we add data.frames rownames as variable to result of split_off with the name specified by rownames. If it is TRUE then name will be .rownames.

Value

split_by returns list of data.frames/split_off returns data.frame

See Also

split, compute, calculate, do_repeat, where

Examples

# example from base R 'split'
data(airquality)
airquality2 = airquality %>%
  split_by(Month) %>%
  compute({
    Ozone_zscore = scale(Ozone)
  }) %>%
  split_off()
head(airquality2)

# usage of 'groups', 'rownames'
data(mtcars)
# add labels to dataset
mtcars %>%
  apply_labels(mpg = "Miles/(US) gallon",
              disp = "Displacement (cu.in.)",
              wt = "Weight",
              hp = "Gross horsepower",
              vs = "Engine",
              am = num_lab("%

  0 V-engine
  1 Straight engine
"),

  am = "Transmission",
  am = num_lab("%

  0 Automatic
  1 Manual
")

) %>%
split_by(am, vs) %>%
use_labels({
  res = lm(mpg ~ hp + disp + wt)
  cbind(Coef. = coef(res), confint(res))
}) %>%
split_off(groups = TRUE, rownames = "variable")

---

### split_labels

**Split character vector to matrix/split columns in data.frame**

**Description**

split_labels/split_columns are auxiliary functions for post-processing tables resulted from cro/cro_fun and etc. In these tables all labels collapsed in the first column with "|" separator. split_columns split first column into multiple columns with separator (split argument). split_table_to_df split first column of table and column names. Result of this operation is data.frame with character columns.

**Usage**

split_labels(
  x,
  remove_repeated = TRUE,
  split = "\|",
  fixed = FALSE,
  perl = FALSE
split_labels

)

split_columns(
    data,
    columns = 1,
    remove_repeated = TRUE,
    split = "\|",
    fixed = FALSE,
    perl = FALSE
)

split_table_to_df(
    data,
    digits = get_expss_digits(),
    remove_repeated = TRUE,
    split = "\|",
    fixed = FALSE,
    perl = FALSE
)

make_subheadings(data, number_of_columns = 1)

Arguments

x character vector which will be split
remove_repeated logical. Default is TRUE. Should we remove repeated labels?
split character vector (or object which can be coerced to such) containing regular
    expression(s) (unless fixed = TRUE) to use for splitting.
fixed logical. If TRUE match split exactly, otherwise use regular expressions. Has
    priority over perl.
perl logical. Should Perl-compatible regexps be used?
data data.frame vector which will be split
columns character/numeric/logical columns in the data.frame data which should be split
digits numeric. How many digits after decimal point should be left in split_table_to_df?
number_of_columns integer. Number of columns from row labels which will be used as subheadings
    in table.

Value

split_labels returns character matrix, split_columns returns data.frame with columns replaced
    by possibly multiple columns with split labels. split_table_to_df returns data.frame with char-
    acter columns.

See Also

strsplit
Examples

data(mtcars)

# apply labels
mtcars = apply_labels(mtcars,
cyl = "Number of cylinders",
vs = "Engine",
"Straight engine" = 1),
am = "Transmission",
am = c(automatic = 0,
manual=1),
gear = "Number of forward gears",
carb = "Number of carburetors"
)

# all row labels in the first column
tabl = mtcars %>%
calculate(cro_cpct(list(cyl, gear, carb), list(total(), vs, am)))

tabl # without subheadings
make_subheadings(tabl) # with subheadings

split_labels(tabl[[1]])
split_labels(colnames(tabl))

# replace first column with new columns
split_columns(tabl) # remove repeated
split_columns(tabl, remove_repeated = FALSE)

split_columns(tabl)

split_table_to_df(tabl)

split_table_to_df(tabl)

sum_row

Compute sum/mean/sd/median/max/min/custom function on rows/columns

Description

These functions are intended for usage inside compute, and do_if. sum/mean/sd/median/max/min by default omits NA. any_in_* checks existence of any TRUE in each row/column. It is equivalent of any applied to each row/column. all_in_* is equivalent of all applied to each row/column.
sum_row

Usage

sum_row(..., na.rm = TRUE)
sum_col(..., na.rm = TRUE)
mean_row(..., na.rm = TRUE)
mean_col(..., na.rm = TRUE)
sd_row(..., na.rm = TRUE)
sd_col(..., na.rm = TRUE)
median_row(..., na.rm = TRUE)
median_col(..., na.rm = TRUE)
max_row(..., na.rm = TRUE)
max_col(..., na.rm = TRUE)
min_row(..., na.rm = TRUE)
min_col(..., na.rm = TRUE)
apply_row(fun, ...)
apply_col(fun, ...)
any_in_row(..., na.rm = TRUE)
any_in_col(..., na.rm = TRUE)
all_in_row(..., na.rm = TRUE)
all_in_col(..., na.rm = TRUE)

Arguments

... data. Vectors, matrices, data.frames, list. Shorter arguments will be recycled.
na.rm logical. Contrary to the base ‘sum’ it is TRUE by default. Should missing values
       (including NaN) be removed?
fun custom function that will be applied to ...

Value

All functions except apply_* return numeric vector of length equals the number of argument
columns/rows. Value of apply_* depends on supplied fun function.
Tables

Functions for custom tables construction

Description

Table construction consists of at least three functions chained with magrittr pipe operator. At first we need to specify variables for which statistics will be computed with `tab_cells`. Secondary, we calculate statistics with one of `tab_stat_*` functions. And last, we finalize table creation with `tab_pivot`: `dataset %>% tab_cells(variable) %>% tab_stat_cases() %>% tab_pivot()`. After that we can optionally sort table with `tab_sort_asc`, drop empty rows/columns with `drop_rc` and transpose with `tab_transpose`. Generally, table is just a data.frame so we can use arbitrary operations on it. Statistic is always calculated with the last cell, column/row variables, weight, missing values and subgroup. To define new cell/column/row variables we can call appropriate function one more time. `tab_pivot` defines how we combine different statistics and where statistic labels will appear - inside/outside rows/columns. See examples. For significance testing see `significance`.

See Also

compute, do_if, %to%, count_if, sum_if, mean_if, median_if, sd_if, min_if, max_if

Examples

```r
iris = compute(iris, {
  new_median = median_row(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width)
  new_mean = mean_row(Sepal.Length, Sepal.Width, Petal.Length, Petal.Width)
})

dfs = data.frame(
  test = 1:5,
  aa = rep(10, 5),
  b_ = rep(20, 5),
  b_1 = rep(11, 5),
  b_2 = rep(12, 5),
  b_4 = rep(14, 5),
  b_5 = rep(15, 5)
)

# calculate sum of b* variables
compute(dfs, {
  b_total = sum_row(b_, b_1 %to% b_5)
})

# conditional modification
do_if(dfs, test %in% 2:4, {
  b_total = sum_row(b_, b_1 %to% b_5)
})
```
Usage

```r
tab_cols(data, ...)

tab_cells(data, ...)

tab_rows(data, ...)

tab_weight(data, weight = NULL)

tab_mis_val(data, ...)

tab_total_label(data, ...)

tab_total_statistic(data, ...)

tab_total_row_position(data, total_row_position = c("below", "above", "none"))

tab_subgroup(data, subgroup = NULL)

tab_row_label(data, ..., label = NULL)

tab_stat_fun(data, ..., label = NULL, unsafe = FALSE)

tab_stat_mean_sd_n(
  data,
  weighted_valid_n = FALSE,
  labels = c("Mean", "Std. dev.", ifelse(weighted_valid_n, "Valid N", "Unw. valid N")),
  label = NULL
)

tab_stat_mean(data, label = "Mean")

tab_stat_median(data, label = "Median")

tab_stat_se(data, label = "S. E.")

tab_stat_sum(data, label = "Sum")

tab_stat_min(data, label = "Min.")

tab_stat_max(data, label = "Max.")

tab_stat_sd(data, label = "Std. dev.")

tab_stat_valid_n(data, label = "Valid N")

tab_stat_unweighted_valid_n(data, label = "Unw. valid N")
```
tab_stat_fun_df(data, ..., label = NULL, unsafe = FALSE)

```
tab_stat_cases(  
data,  
total_label = NULL,  
total_statistic = "u_cases",  
total_row_position = c("below", "above", "none"),  
label = NULL  
)

tab_stat_cpct(  
data,  
total_label = NULL,  
total_statistic = "u_cases",  
total_row_position = c("below", "above", "none"),  
label = NULL  
)

tab_stat_cpct_responses(  
data,  
total_label = NULL,  
total_statistic = "u_responses",  
total_row_position = c("below", "above", "none"),  
label = NULL  
)

tab_stat_tpct(  
data,  
total_label = NULL,  
total_statistic = "u_cases",  
total_row_position = c("below", "above", "none"),  
label = NULL  
)

tab_stat_rpct(  
data,  
total_label = NULL,  
total_statistic = "u_cases",  
total_row_position = c("below", "above", "none"),  
label = NULL  
)

tab_last_vstack(  
data,  
stat_position = c("outside_rows", "inside_rows"),  
stat_label = c("inside", "outside"),  
label = NULL  
)
tab_last_hstack(
  data,
  stat_position = c("outside_columns", "inside_columns"),
  stat_label = c("inside", "outside"),
  label = NULL
)

tab_pivot(
  data,
  stat_position = c("outside_rows", "inside_rows", "outside_columns", "inside_columns"),
  stat_label = c("inside", "outside")
)

tab_transpose(data)

tab_caption(data, ...)

Arguments

data data.frame/intermediate_table
...
weight numeric vector in tab_weight. Cases with NA's, negative and zero weights are removed before calculations.
total_row_position
  Position of total row in the resulting table. Can be one of "below", "above", "none".
subgroup logical vector in tab_subgroup. You can specify subgroup on which table will be computed.
label character. Label for the statistic in the tab_stat_*.
unsafe logical If TRUE than fun will be evaluated as is. It can lead to significant increase in the performance. But there are some limitations. For tab_stat_fun it means that your function fun should return vector of length one. Also there will be no attempts to make labels for statistic. For tab_stat_fun_df your function should return vector of length one or list/data.frame (optionally with 'row_labels' element - statistic labels). If unsafe is TRUE then further arguments (…) for fun will be ignored.
weighted_valid_n
  logical. Sould we show weighted valid N in tab_stat_mean_sd_n? By default it is FALSE.
labels character vector of length 3. Labels for mean, standard deviation and valid N in tab_stat_mean_sd_n.
total_label
  By default "#Total". You can provide several names - each name for each total statistics.
total_statistic

By default it is "u_cases" (unweighted cases). Possible values are "u_cases", "u_responses", "u_cpct", "u_rpct", "w_cases", "w_responses", "w_cpct", "w_rpct", "w_tpct". "u_" means unweighted statistics and "w_" means weighted statistics.

stat_position

character one of the values "outside_rows", "inside_rows", "outside_columns" or "inside_columns". It defines how we will combine statistics in the table.

stat_label

character one of the values "inside" or "outside". Where will be placed labels for the statistics relative to column names/row labels? See examples.

Details

- `tab_cells` variables on which percentage/cases/summary functions will be computed. Use `mrset/mdset` for multiple-response variables.
- `tab_cols` optional variables which breaks table by columns. Use `mrset/mdset` for multiple-response variables.
- `tab_rows` optional variables which breaks table by rows. Use `mrset/mdset` for multiple-response variables.
- `tab_weight` optional weight for the statistic.
- `tab_mis_val` optional missing values for the statistic. It will be applied on variables specified by `tab_cells`. It works in the same manner as `na_if`.
- `tab_subgroup` optional logical vector/expression which specify subset of data for table.
- `tab_row_label` Add to table empty row with specified row labels. It is useful for making section headings and etc.
- `tab_total_row_position` Default value for `total_row_position` argument in `tab_stat_cases` and etc. Can be one of "below", "above", "none".
- `tab_total_label` Default value for `total_label` argument in `tab_stat_cases` and etc. You can provide several names - each name for each total statistics.
- `tab_total_statistic` Default value for `total_statistic` argument in `tab_stat_cases` and etc. You can provide several values. Possible values are "u_cases", "u_responses", "u_cpct", "u_rpct", "w_cases", "w_responses", "w_cpct", "w_rpct", "w_tpct". "u_" means unweighted statistics and "w_" means weighted statistics.
- `tab_stat_fun`, `tab_stat_fun_df` `tab_stat_fun` applies function on each variable in cells separately, `tab_stat_fun_df` gives to function each data.frame in cells as a whole `data.table` with all names converted to variable labels (if labels exists). So it is not recommended to rely on original variables names in your `fun`. For details see `cro_fun`. You can provide several functions as arguments. They will be combined as with `combine_functions`. So you can use method argument. For details see documentation for `combine_functions`.
- `tab_stat_cases` calculate counts.
- `tab_stat_cpct`, `tab_stat_cpct_responses` calculate column percent. These functions give different results only for multiple response variables. For `tab_stat_cpct` base of percent is number of valid cases. Case is considered as valid if it has at least one non-NA value. So for multiple response variables sum of percent may be greater than 100. For `tab_stat_cpct_responses` base of percent is number of valid responses. Multiple response variables can have several responses for single case. Sum of percent of `tab_stat_cpct_responses` always equals to 100%.
• tab_stat_rpct calculate row percent. Base for percent is number of valid cases.
• tab_stat_tpct calculate table percent. Base for percent is number of valid cases.
• tab_stat_mean, tab_stat_median, tab_stat_se, tab_stat_sum, tab_stat_min, tab_stat_max,
  tab_stat_sd, tab_stat_valid_n, tab_stat_unweighted_valid_n different summary statistics. NA's are always omitted.
• tab_pivot finalize table creation and define how different tab_stat_* will be combined
• tab_caption set caption on the table. Should be used after the tab_pivot.
• tab_transpose transpose final table after tab_pivot or last statistic.

Value

All of these functions return object of class intermediate_table except tab_pivot which returns
final result - object of class etable. Basically it's a data.frame but class is needed for custom
methods.

See Also

fre, cro, cro_fun, tab_sort_asc, drop_empty_rows, significance.

Examples

data(mtcars)
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (1000 lbs)",
  qsec = "1/4 mile time",
  vs = "Engine",
  am = "Transmission",
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

# some examples from 'cro'
# simple example - generally with 'cro' it can be made with less typing
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(vs) %>%
  tab_stat_cpct() %>%
  tab_pivot()

# split rows
mtcars %>%
  tab_cells(cyl) %>
tab_cols(vs) %>%
  tab_rows(am) %>%
  tab_stat_cpct() %>%
  tab_pivot()

# multiple banners
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(total(), vs, am) %>%
  tab_stat_cpct() %>%
  tab_pivot()

# nested banners
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(total(), vs %nest% am) %>%
  tab_stat_cpct() %>%
  tab_pivot()

# summary statistics
mtcars %>%
  tab_cells(mpg, disp, hp, wt, qsec) %>%
  tab_cols(am) %>%
  tab_stat_fun(Mean = w_mean, "Std. dev." = w_sd, "Valid N" = w_n) %>%
  tab_pivot()

# summary statistics - labels in columns
mtcars %>%
  tab_cells(mpg, disp, hp, wt, qsec) %>%
  tab_cols(am) %>%
  tab_stat_fun(Mean = w_mean, "Std. dev." = w_sd, "Valid N" = w_n, method = list) %>%
  tab_pivot()

# subgroup with dropping empty columns
mtcars %>%
  tab_subgroup(am == 0) %>%
  tab_cells(cyl) %>%
  tab_cols(total(), vs %nest% am) %>%
  tab_stat_cpct() %>%
  tab_pivot() %>%
  drop_empty_columns()

# total position at the top of the table
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(total(), vs) %>%
  tab_rows(am) %>%
  tab_stat_cpct(total_row_position = "above",
                total_label = c("number of cases", "row %"),
                total_statistic = c("u_cases", "u_rpct")) %>%
  tab_pivot()

# this example cannot be made easily with 'cro'
mtcars %>%
  tab_cells(am) %>%
  tab_cols(total(), vs) %>%
  tab_total_row_position("none") %>%
  tab_stat_cpct(label = "col %") %>%
  tab_stat_rpct(label = "row %") %>%
  tab_stat_tpct(label = "table %") %>%
  tab_pivot(stat_position = "inside_rows")

# statistic labels inside columns
mtcars %>%
  tab_cells(am) %>%
  tab_cols(total(), vs) %>%
  tab_total_row_position("none") %>%
  tab_stat_cpct(label = "col %") %>%
  tab_stat_rpct(label = "row %") %>%
  tab_stat_tpct(label = "table %") %>%
  tab_pivot(stat_position = "inside_columns")

# stacked statistics
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(total(), am) %>%
  tab_stat_mean() %>%
  tab_stat_se() %>%
  tab_stat_valid_n() %>%
  tab_stat_cpct() %>%
  tab_pivot()

# stacked statistics with section headings
mtcars %>%
  tab_cells(cyl) %>%
  tab_cols(total(), am) %>%
  tab_row_label("#Summary statistics") %>%
  tab_stat_mean() %>%
  tab_stat_se() %>%
  tab_stat_valid_n() %>%
  tab_row_label("#Column percent") %>%
  tab_stat_cpct() %>%
  tab_pivot()

# stacked statistics with different variables
mtcars %>%
  tab_cols(total(), am) %>%
  tab_cells(mpg, hp, qsec) %>%
  tab_stat_mean() %>%
  tab_cells(cyl, carb) %>%
  tab_stat_cpct() %>%
  tab_pivot()

# stacked statistics - label position outside row labels
mtcars %>%
  tab_cells(cyl) %>%
```r
# example from 'cro_fun_df' - linear regression by groups with sorting
mtcars %>%
  tab_cells(sheet(mpg, disp, hp, wt, qsec)) %>%
  tab_cols(total(), am) %>%
  tab_stat_fun_df(
    function(x){
      frm = reformulate(".", response = as.name(names(x)[1]))
      model = lm(frm, data = x)
      sheet('Coef.' = coef(model),
            confint(model)
    }
  ) %>%
  tab_pivot() %>%
  tab_sort_desc()

# multiple-response variables and weight
data(product_test)
codeframe_likes = num_lab("1 Liked everything
2 Disliked everything
3 Chocolate
4 Appearance
5 Taste
6 Stuffing
7 Nuts
8 Consistency
98 Other
99 Hard to answer
")

set.seed(1)
product_test = compute(product_test, {
  # recode age by groups
  age_cat = recode(s2a, lo %thru% 25 ~ 1, lo %thru% hi ~ 2)
  var_lab(age_cat) = "Age"
  val_lab(age_cat) = c("18 - 25" = 1, "26 - 35" = 2)

  var_lab(a1_1) = "Likes. VSX123"
  var_lab(b1_1) = "Likes. SDF456"
  val_lab(a1_1) = codeframe_likes
  val_lab(b1_1) = codeframe_likes

  wgt = runif(.N, 0.25, 4)
  wgt = wgt/sum(wgt)*.N
```

product_test %>%
  tab_cells(mrset(a1_1 %to% a1_6), mrset(b1_1 %to% b1_6)) %>%
  tab_cols(total(), age_cat) %>%
  tab_weight(wgt) %>%
  tab_stat_cpct() %>%
  tab_sort_desc() %>%
  tab_pivot()

# trick to place cell variables labels inside columns
# useful to compare two variables
# '|' is needed to prevent automatic labels creation from argument
# alternatively we can use list(...) to avoid this
product_test %>%
  tab_cols(total(), age_cat) %>%
  tab_weight(wgt) %>%
  tab_cells("|") = unvr(mrset(a1_1 %to% a1_6))) %>%
  tab_stat_cpct(label = var_lab(a1_1)) %>%
  tab_cells("|") = unvr(mrset(b1_1 %to% b1_6))) %>%
  tab_stat_cpct(label = var_lab(b1_1)) %>%
  tab_pivot(stat_position = "inside_columns")

# if you need standard evaluation, use 'vars'
tables = mtcars %>%
  tab_cols(total(), am %nest% vs)
for(each in c("mpg", "disp", "hp", "qsec")){
  tables = tables %>% tab_cells(vars(each)) %>%
    tab_stat_fun(Mean = w_mean, "Std. dev." = w_sd, "Valid N" = w_n)
}
tables %>% tab_pivot()

---

**tab_significance_options**

*Mark significant differences between columns in the table*

**Description**

- **significance_cpct** conducts z-tests between column percent in the result of `cro_cpct`. Results are calculated with the same formula as in `prop.test` without continuity correction.

- **significance_means** conducts t-tests between column means in the result of `cro_mean_sd_n`. Results are calculated with the same formula as in `t.test`.

- **significance_cases** conducts chi-squared tests on the subtable of table with counts in the result of `cro_cases`. Results are calculated with the same formula as in `chisq.test`.

- **significance_cell_chisq** compute cell chi-square test on table with column percent. The cell chi-square test looks at each table cell and tests whether it is significantly different from its expected value in the overall table. For example, if it is thought that variations in political
opinions might depend on the respondent’s age, this test can be used to detect which cells contribute significantly to that dependence. Unlike the chi-square test (significance_cases), which is carried out on a whole set of rows and columns, the cell chi-square test is carried out independently on each table cell. Although the significance level of the cell chi-square test is accurate for any given cell, the cell tests cannot be used instead of the chi-square test carried out on the overall table. Their purpose is simply to point to the parts of the table where dependencies between row and column categories may exist.

For significance_cpct and significance_means there are three type of comparisons which can be conducted simultaneously (argument compare_type):

- **subtable** provide comparisons between all columns inside each subtable.
- **previous_column** is a comparison of each column of the subtable with the previous column. It is useful if columns are periods or survey waves.
- **first_column** provides comparison the table first column with all other columns in the table. **adjusted_first_column** is also comparison with the first column but with adjustment for common base. It is useful if the first column is total column and other columns are subgroups of this total. Adjustments are made according to algorithm in IBM SPSS Statistics Algorithms v20, p. 263. Note that with these adjustments t-tests between means are made with equal variance assumed (as with var_equal = TRUE).

By now there are no adjustments for multiple-response variables (results of mrset) in the table columns so significance tests are rather approximate for such cases. Also, there are functions for the significance testing in the sequence of custom tables calculations (see tables):

- **tab_last_sig_cpct**, **tab_last_sig_means** and **tab_last_sig_cpct** make the same tests as their analogs mentioned above. It is recommended to use them after appropriate statistic function: **tab_stat_cpct**, **tab_stat_mean_sd_n** and **tab_stat_cases**.
- **tab_significance_options** With this function we can set significance options for the entire custom table creation sequence.
- **tab_last_addsig_labels** This function applies add_sig_labels to the last calculated table - it adds labels (letters by default) for significance to columns header. It may be useful if you want to combine a table with significance with table without it.
- **tab_last_round** This function rounds numeric columns in the last calculated table to specified number of digits. It is sometimes needed if you want to combine table with significance with table without it.

**Usage**

```r
tab_significance_options(
  data,
  sig_level = 0.05,
  min_base = 2,
  delta_cpct = 0,
  delta_means = 0,
  correct = TRUE,
  compare_type = "subtable",
  bonferroni = FALSE,
  subtable_marks = "greater",
)```
inequality_sign = "both" %in% subtable_marks,
sig_labels = LETTERS,
sig_labels_previous_column = c("v", "^"),
sig_labels_first_column = c("-", "+"),
sig_labels_chisq = c("<", ">"),
keep = c("percent", "cases", "means", "sd", "bases"),
row_margin = c("auto", "sum_row", "first_column"),
total_marker = "+",
total_row = 1,
digits = get_expss_digits(),
na_as_zero = FALSE,
var_equal = FALSE,
mode = c("replace", "append")
)

tab_last_sig_cpct(
data,
sig_level = 0.05,
delta_cpct = 0,
min_base = 2,
compare_type = "subtable",
bonferroni = FALSE,
subtable_marks = c("greater", "both", "less"),
inequality_sign = "both" %in% subtable_marks,
sig_labels = LETTERS,
sig_labels_previous_column = c("v", "^"),
sig_labels_first_column = c("-", "+"),
keep = c("percent", "bases"),
na_as_zero = FALSE,
total_marker = "+",
total_row = 1,
digits = get_expss_digits(),
mode = c("replace", "append"),
label = NULL
)

tab_last_sig_means(
data,
sig_level = 0.05,
delta_means = 0,
min_base = 2,
compare_type = "subtable",
bonferroni = FALSE,
subtable_marks = c("greater", "both", "less"),
inequality_sign = "both" %in% subtable_marks,
sig_labels = LETTERS,
sig_labels_previous_column = c("v", "^"),
sig_labels_first_column = c("-", "+"),
tab_significance_options

```r
keep = c("means", "sd", "bases"),
var_equal = FALSE,
digits = get_expss_digits(),
mode = c("replace", "append"),
label = NULL)

tab_last_sig_cases(
data,
sig_level = 0.05,
min_base = 2,
correct = TRUE,
keep = c("cases", "bases"),
total_marker = ",
total_row = 1,
digits = get_expss_digits(),
mode = c("replace", "append"),
label = NULL)

tab_last_sig_cell_chisq(
data,
sig_level = 0.05,
min_base = 2,
subtable_marks = c("both", "greater", "less"),
sig_labels_chisq = c("<", ",
correct = TRUE,
keep = c("percent", "bases", "none"),
row_margin = c("auto", "sum_row", "first_column"),
total_marker = ",
total_row = 1,
total_column_marker = ",
digits = get_expss_digits(),
mode = c("replace", "append"),
label = NULL)

tab_last_round(data, digits = get_expss_digits())

tab_last_add_sig_labels(data, sig_labels = LETTERS)

significance_cases(
  x,
sig_level = 0.05,
min_base = 2,
correct = TRUE,
keep = c("cases", "bases"),
total_marker = ",
```
total_row = 1,
    digits = get_expss_digits()
)

significance_cell_chisq(
    x,
    sig_level = 0.05,
    min_base = 2,
    subtable_marks = c("both", "greater", "less"),
    sig_labels_chisq = c("<", ">"),
    correct = TRUE,
    keep = c("percent", "bases", "none"),
    row_margin = c("auto", "sum_row", "first_column"),
    total_marker = "#",
    total_row = 1,
    total_column_marker = "#",
    digits = get_expss_digits()
)

cell_chisq(cases_matrix, row_base, col_base, total_base, correct)

significance_cpct(
    x,
    sig_level = 0.05,
    delta_cpct = 0,
    min_base = 2,
    compare_type = "subtable",
    bonferroni = FALSE,
    subtable_marks = c("greater", "both", "less"),
    inequality_sign = "both" %in% subtable_marks,
    sig_labels = LETTERS,
    sig_labels_previous_column = c("v", "^"),
    sig_labels_first_column = c("-", "+"),
    keep = c("percent", "bases"),
    na_as_zero = FALSE,
    total_marker = "#",
    total_row = 1,
    digits = get_expss_digits()
)

add_sig_labels(x, sig_labels = LETTERS)

significance_means(
    x,
    sig_level = 0.05,
    delta_means = 0,
    min_base = 2,
    compare_type = "subtable",
    correct = TRUE,
bonferroni = FALSE,
subtable_marks = c("greater", "both", "less"),
inequality_sign = "both" %in% subtable_marks,
sig_labels = LETTERS,
sig_labels_previous_column = c("v", "^"),
sig_labels_first_column = c("-", "+"),
keep = c("means", "sd", "bases"),
var_equal = FALSE,
digits = get_expss_digits()
)

Arguments

data       data.frame/intermediate_table for tab_* functions.
sig_level  numeric. Significance level - by default it equals to 0.05.
min_base   numeric. Significance test will be conducted if both columns have bases greater or equal to min_base. By default, it equals to 2.
delta_cpct numeric. Minimal delta between percent for which we mark significant differences (in percent points) - by default it equals to zero. Note that, for example, for minimal 5 percent point difference delta_cpct should be equals 5, not 0.05.
delta_means numeric. Minimal delta between means for which we mark significant differences - by default it equals to zero.
correct    logical indicating whether to apply continuity correction when computing the test statistic for 2 by 2 tables. Only for significance_cases and significance_cell_chisq. For details see chisq.test. TRUE by default.
compare_type Type of compare between columns. By default, it is subtable - comparisons will be conducted between columns of each subtable. Other possible values are: first_column, adjusted_first_column and previous_column. We can conduct several tests simultaneously.
bonferroni logical. FALSE by default. Should we use Bonferroni adjustment by the number of comparisons in each row?
subtable_marks character. One of "greater", "both" or "less". By default we mark only values which are significantly greater than some other columns. For significance_cell_chisq default is "both". We can change this behavior by setting an argument to less or both.
inequality_sign logical. FALSE if subtable_marks is "less" or "greater". Should we show > or < before significance marks of subtable comparisons.
sig_labels character vector. Labels for marking differences between columns of subtable.
sig_labels_previous_column a character vector with two elements. Labels for marking a difference with the previous column. First mark means 'lower' (by default it is v) and the second means greater (^).
sig_labels_first_column a character vector with two elements. Labels for marking a difference with the first column of the table. First mark means 'lower' (by default it is -) and the second means 'greater' (+).
sig_labels_chisq
a character vector with two labels for marking a difference with row margin of the table. First mark means 'lower' (by default it is <) and the second means 'greater' (>). Only for significance_cell_chisq.

keep
character. One or more from "percent", "cases", "means", "bases", "sd" or "none". This argument determines which statistics will remain in the table after significance marking.

row_margin
character. One of values "auto" (default), "sum_row", or "first_column". If it is "auto" we try to find total column in the subtable by total_column_marker. If the search is failed, we use the sum of each rows as row total. With "sum_row" option we always sum each row to get margin. Note that in this case result for multiple response variables in banners may be incorrect. With "first_column" option we use table first column as row margin for all subtables. In this case result for the subtables with incomplete bases may be incorrect. Only for significance_cell_chisq.

total_marker
character. Total rows mark in the table. "#" by default.

total_row
integer/character. In the case of the several totals per subtable it is a number or name of total row for the significance calculation.

digits
an integer indicating how much digits after decimal separator will be shown in the final table.

na_as_zero
logical. FALSE by default. Should we treat NA's as zero cases?

var_equal
a logical variable indicating whether to treat the two variances as being equal. For details see t.test.

mode
character. One of replace (default) or append. In the first case the previous result in the sequence of table calculation will be replaced with result of significance testing. In the second case result of the significance testing will be appended to sequence of table calculation.

label
character. Label for the statistic in the tab_*_. Ignored if the mode is equals to replace.

total_column_marker
character. Mark for total columns in the subtables. "#" by default.

x
table (class etable): result of cro_cpt with proportions and bases for significance_cpct, result of cro_mean_sd_n with means, standard deviations and valid N for significance_means, and result of cro_cases with counts and bases for significance_cases.

cases_matrix
numeric matrix with counts size R*C

cases_matrix
numeric vector with row bases, length R

col_base
numeric vector with col bases, length C

total_base
numeric single value, total base

Value

tab_last_* functions return objects of class intermediate_table. Use tab_pivot to get the final result - etable object. Other functions return etable object with significant differences.
See Also
cro_cpct, cro_cases, cro_mean_sd_n, tables, compare_proportions, compare_means, prop.test, t.test, chisq.test

Examples

data(mtcars)
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
cyl = "Number of cylinders",
disp = "Displacement (cu.in.)",
hp = "Gross horsepower",
trans = "Rear axle ratio",
wgt = "Weight (lb/1000)",
qsec = "1/4 mile time",
vs = "Engine",
vs = c("V-engine" = 0,
  "Straight engine" = 1),
am = "Transmission",
am = c("Automatic" = 0,
  "Manual" = 1),
gear = "Number of forward gears",
carb = "Number of carburetors"
)

mtcars_table = calculate(mtcars,
  cro_cpct(list(cyl, gear),
    list(total(), vs, am))
)

significance_cpct(mtcars_table)
## Not run:
# comparison with first column
significance_cpct(mtcars_table, compare_type = "first_column")

# comparison with first column and inside subtable
significance_cpct(mtcars_table,
  compare_type = c("first_column", "subtable"))

# only significance marks
significance_cpct(mtcars_table, keep = "none")

# means
mtcars_means = calculate(mtcars,
  cro_mean_sd_n(list(mpg, wt, hp),
    list(total(), vs, cyl))
)

significance_means(mtcars_means)
# mark values which are less and greater
significance_means(mtcars_means, subtable_marks = "both")
# chi-squared test
mtcars_cases = calculate(mtcars,
cro_cases(list(cyl, gear),
  list(total(), vs, am))
)
significance_cases(mtcars_cases)

# cell chi-squared test
# increase number of cases to avoid warning about chi-square approximation
mtcars2 = add_rows(mtcars, mtcars, mtcars)
tbl = calc_cro_cpct(mtcars2, gear, am)
significance_cell_chisq(tbl)

# table with multiple variables
tbl = calc_cro_cpct(mtcars2, list(gear, cyl), list(total(), am, vs))
significance_cell_chisq(tbl, sig_level = .0001)

# custom tables with significance
mtcars %>%
  tab_significance_options(subtable_marks = "both") %>%
  tab_cells(mpg, hp) %>%
  tab_cols(total(), vs, am) %>%
  tab_stat_mean_sd_n() %>%
  tab_last_sig_means(keep = "means") %>%
  tab_cells(cyl, gear) %>%
  tab_stat_cpct() %>%
  tab_last_sig_cpct() %>%
  tab_pivot()

# Overcomplicated examples - we move significance marks to
# separate columns. Columns with statistics remain numeric
mtcars %>%
  tab_significance_options(keep = "none",
    sig_labels = NULL,
    subtable_marks = "both",
    mode = "append") %>%
  tab_cols(total(), vs, am) %>%
  tab_cells(mpg, hp) %>%
  tab_stat_mean_sd_n() %>%
  tab_last_sig_means() %>%
  tab_last_hstack("inside_columns") %>%
  tab_cells(cyl, gear) %>%
  tab_stat_cpct() %>%
  tab_last_sig_cpct() %>%
  tab_last_hstack("inside_columns") %>%
  tab_pivot(stat_position = "inside_rows") %>%
  drop_empty_columns()
Description

tab_sort_asc/tab_sort_desc sort tables (usually result of cro/tables) in ascending/descending order between specified rows (by default, it is rows which contain ‘#’ in the first column).

Usage

tab_sort_asc(x, ..., excluded_rows = "#", na.last = FALSE)
tab_sort_desc(x, ..., excluded_rows = "#", na.last = TRUE)

Arguments

x data.frame

... character/numeric or criteria/logical functions (see criteria). Column names/numbers for data.frame/matrix by which object will be sorted. Names at the top-level can be unquoted (non-standard evaluation). For standard evaluation of parameters you can surround them by round brackets. See examples. If this argument is missing then table will be sorted by second column. Usually second column is the first column with numbers in the table (there are row labels in the first column).

excluded_rows character/logical/numeric rows which won’t be sorted. Rows of the table will be sorted between excluded rows. If it is characters then they will be considered as pattern/vector of patterns. Patterns will be matched with Perl-style regular expression with values in the first column of x (see grep, perl = TRUE argument). Rows which have such patterns will be excluded. By default, pattern is "#" because "#" marks totals in the result of cro.

na.last for controlling the treatment of NAs. If TRUE, missing values in the data are put last; if FALSE, they are put first; if NA, they are removed.

Value

sorted table('etable')/data.frame

Examples

data(mtcars)

# apply labels
mtcars = apply_labels(mtcars,
cyl = "Number of cylinders",
vs = "Engine",
vs = c("V-engine" = 0,
   "Straight engine" = 1),
text_to_columns

Make data.frame from text

Description

Convert delimited text lines to data.frame. Blank lines are always skipped, trailing whitespaces are trimmed. You can use comments with "#" inside your text. For details see read.table.

Usage

text_to_columns(
  text,
  header = TRUE,
  sep = " ",
  quote = " ",
  dec = ".",
  encoding = "unknown",
  ...
)

text_to_columns_csv(
  text,
  header = TRUE,
  sep = ",",
  quote = ",",
  dec = ".",
  ...)
text_to_columns_csv2(
  text,
  header = TRUE,
  sep = ":",
  quote = ":",
  dec = ",",
  encoding = "unknown",
  ...
)

text_to_columns_tab(
  text,
  header = TRUE,
  sep = "\t",
  quote = "",
  dec = ".",
  encoding = "unknown",
  ...
)

text_to_columns_tab2(
  text,
  header = TRUE,
  sep = "\t",
  quote = "",
  dec = ",",
  encoding = "unknown",
  ...
)

Arguments

text character/vector of characters
header a logical value indicating whether the text contains the names of the variables as its first line.
sep the field separator character. Values on each line of the file are separated by this character. If sep = "" (the default for text_to_columns) the separator is 'white space', that is one or more spaces, tabs, newlines or carriage returns.
quote the set of quoting characters. To disable quoting altogether, use quote = ":".
dec the character used in the file for decimal points.
encoding encoding to be assumed for input strings. It is used to mark character strings as known to be in Latin-1 or UTF-8 (see read.table).
...
  further parameters which will be passed to read.table.
unlab

Value
data.frame

Examples
text_to_columns("# simple data.frame
  a b c
  1 2.5 a
  4 5.5 b
  7 8.5 c
")

unlab Drop variable label and value labels

Description
unlab returns variable x without variable labels and value labels

Usage
unlab(x)
drop_all_labels(x)

Arguments
x Variable(s). Vector/data.frame/list.

Value
unlab returns original variable x without variable label, value labels and class.

See Also
unvr unvl

Examples
raw_var = rep(1:2,5)
var_with_lab = set_var_lab(raw_var,"Income")
val_lab(var_with_lab) = c("Low"=1,"High"=2)
identical(raw_var,unlab(var_with_lab)) # should be TRUE
values2labels

Replace vector/matrix/data.frame/list values with corresponding value labels.

Description

values2labels replaces vector/matrix/data.frame/list values with corresponding value labels. If there are no labels for some values they are converted to characters in most cases. If there are no labels at all for variable it remains unchanged. v2l is just shortcut to values2labels.

Usage

values2labels(x)

v2l(x)

Arguments

x vector/matrix/data.frame/list

Value

Object of the same form as x but with value labels instead of values.

See Also

names2labels, val_lab, var_lab

Examples

data(mtcars)
mtcars = modify(mtcars,
    var_lab(mpg) = NULL
    var_lab(am) = c(" automatic" = 0, " manual" = 1)
)

summary(lm(mpg ~ ., data = values2labels(mtcars[,c("mpg","am")]))
val_lab

Set or get value labels

Description

These functions set/get/drop value labels. Duplicated values are not allowed. If argument x is data.frame or list then labels applied to all elements of data.frame/list. To drop value labels, use val_lab(var) <- NULL or unvl(var). make_labels converts text from the form that usually used in questionnaires to named vector. For variable labels see var_lab. For working with entire data.frame see apply_labels.

- val_lab returns value labels or NULL if labels doesn’t exist.
- val_lab<- set value labels.
- set_val_lab returns variable with value labels.
- add_val_lab<- add value labels to already existing value labels.
- unvl drops value labels.
- make_labels makes named vector from text for usage as value labels.
- num_lab, lab_num and autonum are shortcuts for make_labels with code_position 'left', 'right' and 'autonum' accordingly.

Usage

val_lab(x)

val_lab(x) <- value

set_val_lab(x, value, add = FALSE)

add_val_lab(x, value)

add_val_lab(x) <- value

unvl(x)

drop_val_labs(x)

make_labels(text, code_position = c("left", "right", "autonum"))

drop_unused_labels(x)

num_lab(text)

lab_num(text)

autonum(text)
Arguments

- **x**: Variable(s). Vector/data.frame/list.
- **value**: Named vector. Names of vector are labels for the appropriate values of variable `x`.
- **add**: Logical. Should we add value labels to old labels or replace it? Default is FALSE - we completely replace old values. If TRUE new value labels will be combined with old value labels.
- **text**: Text that should be converted to named vector.
- **code_position**: Possible values "left", "right" - position of numeric code in text. "autonum" - makes codes by autonumbering lines of text.

Details

Value labels are stored in attribute "labels" (attr(x,"labels"). We set variable class to "labelled" for preserving labels from dropping during some operations (such as c and `[]`).

Value

`val_lab` return value labels (named vector). If labels doesn't exist it return NULL. `val_lab<-` and `set_val_lab` return variable (vector x) of class "labelled" with attribute "labels" which contains value labels. `make_labels` return named vector for usage as value labels.

Examples

```r
# toy example
set.seed(123)
# score - evaluation of tested product

score = sample(-1:1,20,replace = TRUE)
var_lab(score) = "Evaluation of tested brand"
val_lab(score) = c("Dislike it" = -1,
                   "So-so" = 0,
                   "Like it" = 1
               )

# frequency of product scores
fre(score)

# brands - multiple response question
# Which brands do you use during last three months?

brands = as.sheet(t(replicate(20,sample(c(1:5,NA),4,replace = FALSE))))

var_lab(brands) = "Used brands"
val_lab(brands) = make_labels("1 Brand A
2 Brand B
3 Brand C
4 Brand D
5 Brand E")
```
vars

Get variables/range of variables by name/by pattern.

Description

- **vars** returns data.frame with all variables by their names or by criteria (see **criteria**). There is no non-standard evaluation in this function by design so use quotes for names of your variables. This function is intended to get variables by parameter/criteria. The only exception with non-standard evaluation is %to%. You can use %to% inside vars or independently.
- **..p** returns data.frame with all variables which names satisfy supplied perl-style regular expression. Arguments for this function is quoted characters. It is a shortcut for `vars(perl(pattern))`. 
• . . f returns data.frame with all variables which names contain supplied pattern. Arguments for this function can be unquoted. It is a shortcut for `vars(fixed(pattern))`.

• . . t returns data.frame with variables which names are stored in the supplied arguments. Expressions in characters in curly brackets are expanded. See `text_expand`.

• . . [ ] returns data.frame with all variables by their names or by criteria (see `criteria`). Names at the top-level can be unquoted (non-standard evaluation). For standard evaluation of parameters you can surround them by round brackets. If there are several names inside square brackets then each element of list/data.frame from right side will be assigned to appropriate name from left side. You can use `item1 %to% item2` notation to get/create sequence of variables. If there are no arguments inside square brackets than from each item of RHS will be created separate variable in the parent frame. In this case RHS should be named list or data.frame.

• . . $name sets/returns object which name is stored in the variable name. It is convenient wrapper around `get/assign` functions.

• %to% returns range of variables between `e1` and `e2` (similar to SPSS 'to'). `modify`, `modify_if`, `calculate`, `keep`, `except` and `where` support `%to%`.

• `indirect`/`indirect_list` are aliases for `vars/vars_list`.

Functions with word 'list' in name return lists of variables instead of dataframes.

Usage

`vars(...)`

`vars_list(...)`

`indirect(...)`

`indirect_list(...)`

`e1 %to% e2`

`e1 %to_list% e2`

`..`

`..f(...)`

`..p(...)`

`..t(...)`

Arguments

... characters names of variables or criteria/logical functions

`e1` unquoted name of start variable (e. g. a_1)

`e2` unquoted name of start variable (e. g. a_5)
vars

Format
An object of class parameter of length 1.

Value
data.frame/list with variables

See Also
keep, except, do_repeat, compute, calculate, where

Examples

```r
# In data.frame
dfs = data.frame(
  a = rep(1, 5),
  b_1 = rep(11, 5),
  b_2 = rep(12, 5),
  b_3 = rep(13, 5),
  b_4 = rep(14, 5),
  b_5 = rep(15, 5)
)

# calculate sum of b_* variables
compute(dfs, {
  b_total = sum_row(b_1 %to% b_5)
})

# identical result
compute(dfs, {
  b_total = sum_row(..f(b_))
})

compute(dfs, {
  b_total = sum_row(..t("b_{1:5}"))
})

# In global environment
a = rep(10, 5)
a1 = rep(1, 5)
a2 = rep(2, 5)
a3 = rep(3, 5)
a4 = rep(4, 5)
a5 = rep(5, 5)

# identical results
a1 %to% a5
vars(perl("a[0-9]$"))
..[perl("a[0-9]$")]
..p("a[0-9]$")
..t("a{l:5}"を探す
### sum each row

```r
sum_row(a1 %to% a5)
```

### variable substitution

```r
name1 = "a"
name2 = "new_var"
```

### in global environment

```r
..$name1 # give as variable 'a'

..$name2 = ..$name1 * 2 # create variable 'new_var' which is equal to 'a' times 2
```

### inside data.frame

```r
compute(dfs, {
  ..$name2 = ..$name1*2
})
```

```r
compute(dfs, {
  for(name1 in paste0("b_", 1:5)){
    name2 = paste0("new_", name1)
    ..$name2 = ..$name1*2
  }

  rm(name1, name2) # we don't need this variables as columns in 'dfs'
})
```

### square brackets notation - multi-assignment

```r
name1 = paste0("b_", 1:5)
compute(dfs, {
  # round brackets about 'name1' is needed to avoid using it 'as is'
  ..[paste0("new_", name1)] = ..[(name1)]*2
})
```

### the same result

```r
# note the automatic creation of sequence of variables
compute(dfs, {
  ..[new_b_1 %to% new_b_5] = ..[b_1 %to% b_5]*2
})
```

### assignment form of 'recode' on multiple variables

```r
compute(dfs, {
  recode(.[b_1 %to% b_5]) = 13 %thru% hi - 20
})
```

### empty brackets - unboxing of dichotomy.

```r
compute(dfs, {
  [.[] = as.dichotomy(b_1 %to% b_5, prefix = "v_")
})
```

---

**var_lab**

**Set or get variable label**
var_lab

Description

These functions set/get/drop variable labels. For value labels see val_lab. For working with entire data.frame see apply_labels.

- var_lab returns variable label or NULL if label doesn’t exist.
- var_lab<- set variable label.
- set_var_lab returns variable with label.
- unvr drops variable label.
- add_labelled_class Add missing 'labelled' class. This function is needed when you load SPSS data with packages which in some cases don’t set 'labelled' class for variables with labels. For example, haven package doesn’t set 'labelled' class for variables which have variable label but don’t have value labels. Note that to use 'expss' with 'haven' you need to load 'expss' strictly after 'haven' to avoid conflicts.

Usage

```r
var_lab(x, default = NULL)
var_lab(x) <- value
set_var_lab(x, value)
unvr(x)
drop_var_labs(x)
add_labelled_class(x, remove_classes = c("haven_labelled", "spss_labelled", "haven_labelled_spss", "vctrs_vctr")
```

Arguments

- `x` Variable. In the most cases it is numeric vector.
- `default` A character scalar. What we want to get from 'var_lab' if there is no variable label. NULL by default.
- `value` A character scalar - label for the variable x.
- `remove_classes` A character vector of classes which should be removed from the class attribute of the x.

Details

Variable label is stored in attribute "label" (attr(x,"label")). For preserving from dropping this attribute during some operations (such as c) variable class is set to "labelled". There are special methods of subsetting and concatenation for this class. To drop variable label use var_lab(var) <-NULL or unvr(var).
Value

\texttt{var\_lab} return variable label. If label doesn’t exist it return NULL. \texttt{\_lab<-} and \texttt{set\_var\_lab} return variable (vector x) of class "labelled" with attribute "label" which equals submitted value.

Examples

\begin{verbatim}
data(mtcars)
mtcars = modify(mtcars,
   var.lab(mpg) = "Miles/(US) gallon"
   var.lab(cyl) = "Number of cylinders"
   var.lab(disp) = "Displacement (cu.in.)"
   var.lab(hp) = "Gross horsepower"
   var.lab(drat) = "Rear axle ratio"
   var.lab(wt) = "Weight (lb/1000)"
   var.lab(qsec) = "1/4 mile time"
   var.lab(vs) = "V/S"
   var.lab(am) = "Transmission (0 = automatic, 1 = manual)"
   val.lab(am) = c(automatic = 0, manual=1)
   var.lab(gear) = "Number of forward gears"
   var.lab(carb) = "Number of carburetors"
)
fre(mtcars$am)
calculate(mtcars,
cro_mean(list(mpg, disp, hp, qsec), list(total(), am))
)
\end{verbatim}

\texttt{\_lab<-} and \texttt{set\_var\_lab} return variable (vector x) of class "labelled" with attribute "label" which equals submitted value.

Vectors

Infix operations on vectors - \texttt{append}, \texttt{diff}, \texttt{intersection}, \texttt{union}, \texttt{replication}

Description

All these functions except \texttt{\_d\%}, \texttt{\_i\%} preserve names of vectors and don’t remove duplicates.

- \texttt{\_a\%} \texttt{a(ppends)} second argument to the first argument. See also \texttt{append}.
vectors

- `%u%` and `v_union` (n)ite first and second arguments. Remove elements from the second argument which exist in the first argument.
- `%d%` and `v_diff` (iff) second argument from the first argument. Second argument could be a function which returns logical value. In this case elements of the first argument which give TRUE will be removed.
- `%i%` and `v_intersect` (ntersect) first argument and second argument. Second argument could be a function which returns logical value. In this case elements of the first argument which give FALSE will be removed.
- `%e%` and `v_xor` (exclusive OR). Returns elements that contained only in one of arguments.
- `%r%` (epeats) first argument second argument times. See also `rep`.
- `%n_d%` and `n_diff` (ames) (iff) - diff second argument from names of first argument. Second argument could be a function which returns logical value. In this case elements of the first argument which names give TRUE will be removed.
- `%n_i%` and `n_intersect` (ames) (ntersect) - intersect names of the first argument with the second argument. Second argument could be a function which returns logical value. In this case elements of the first argument which names give FALSE will be removed.

For `%d%`, `%i%`, `%n_d%`, `%n_i%` one can use criteria functions. See `criteria` for details.

Usage

\[
e1 %a% e2
\]
\[
v_union(e1, e2)
\]
\[
e1 %u% e2
\]
\[
v_diff(e1, e2)
\]
\[
e1 %d% e2
\]
\[
v_intersect(e1, e2)
\]
\[
e1 %i% e2
\]
\[
v_xor(e1, e2)
\]
\[
e1 %e% e2
\]
\[
e1 %r% e2
\]
\[
n_intersect(e1, e2)
\]
\[
e1 %n_i% e2
\]
\[
n_diff(e1, e2)
\]
\[
e1 %n_d% e2
\]
Arguments

- \( e1 \) vector or data.frame, matrix, list for \( \%n_d\), \( \%n_i\)
- \( e2 \) vector or function for \( \%d\), \( \%i\)

Value

vector or data.frame, matrix, list for \( \%n_d\), \( \%n_i\)

Examples

```
1:4 %a% 5:6  # 1:6
1:4 %a% 4:5  # 1,2,3,4,4,5
1:4 %u% 4:5  # 1,2,3,4,5
1:6 %d% 5:6  # 1:4

# function as criterion
1:6 %d% greater(4) # 1:4
1:4 %i% 4:5  # 4

# with criteria functions
letters %i% (contains("a") | contains("z")) # a, z
letters %i% perl("[a-d]") # a,b,c,d
1:4 %e% 4:5  # 1, 2, 3, 5
1:2 %r% 2    # 1, 2, 1, 2

# %n_i%, %n_d%

# remove column Species
iris %n_d% "Species"

# leave only columns which names start with "Sepal"
iris %n_i% like("Sepal*")

# leave column "Species" and columns which names start with "Sepal"
iris %n_i% ("Species" | like("Sepal*"))
iris %n_i% or("Species", like("Sepal*")) # same result
```
Description

vlookup/vlookup_df function is inspired by VLOOKUP spreadsheet function. It looks for a
lookup_value in the lookup_column of the dict, and then returns values in the same rows from
result_column. add_columns inspired by MATCH FILES (Add variables...) from SPSS Statistics. It works similar to SQL left join but number of cases in the left part always remain the same.
If there are duplicated keys in the dict then error will be raised by default. add_columns is the
same function for default dataset.

Usage

vlookup(lookup_value, dict, result_column = 2, lookup_column = 1)

vlookup_df(lookup_value, dict, result_column = NULL, lookup_column = 1)

add_columns(data, dict, by = NULL, ignore_duplicates = FALSE)

..add_columns(dict, by = NULL, ignore_duplicates = FALSE)

Arguments

lookup_value Vector of looked up values
result_column numeric or character. Resulting columns of dict. There are special values:
'row.names', 'rownames', 'names'. If result_column equals to one of these
special values and dict is matrix/data.frame then row names of dict will be re-
turned. If dict is vector then names of vector will be returned. For vlookup_df
default result_column is NULL and result will be entire rows. For vlookup
default result_column is 2 · for frequent case of dictionary with keys in the first
column and results in the second column.
lookup_column Column of dict in which lookup value will be searched. By default it is the
first column of the dict. There are special values: 'row.names', 'rownames',
'names'. If lookup_column equals to one of these special values and dict is
matrix/data.frame then values will be searched in the row names of dict. If
dict is vector then values will be searched in names of the dict.
data data.frame to be joined with dict.
by character vector or NULL(default) or 1. Names of common variables in the data
and dict by which we will attach dict to data. If it is NULL then common
names will be used. If it is equals to 1 then we will use the first column from
both dataframes. To add columns by different variables on data and dict use a
named vector. For example, by = c("a" = "b") will match data.a to dict.b.
ignore_duplicates logical Should we ignore duplicates in the by variables in the dict? If it is
TRUE than first occurrence of duplicated key will be used.

Value

vlookup always return vector, vlookup_df always returns data.frame. row.names in result of
vlookup_df are not preserved.
Examples

# with data.frame
dict = data.frame(num=1:26, small=letters, cap=LETTERS, stringsAsFactors = FALSE)
rownames(dict) = paste0('rows', 1:26)
identical(vlookup_df(1:3, dict), dict[1:3,]) # should be TRUE
vlookup(c(45,1:3,58), dict, result_column='cap')
vlookup_df(c('z','d','f'), dict, lookup_column = 'small')
vlookup_df(c('rows7', 'rows2', 'rows5'), dict, lookup_column = 'row.names')

# with vector
dict=1:26
names(dict) = letters
vlookup(c(2,4,6), dict, result_column='row.names')

# The same results
vlookup(c(2,4,6), dict, result_column='rownames')
vlookup(c(2,4,6), dict, result_column='names')

# example for 'add_columns' from base 'merge'
authors = sheet(
  surname = c("Tukey", "Venables", "Tierney", "Ripley", "McNeil"),
  nationality = c("US", "Australia", "US", "UK", "Australia"),
  deceased = c("yes", rep("no", 4))
)
books = sheet(
  surname = c("Tukey", "Venables", "Tierney",
  "Ripley", "Ripley", "McNeil", "R Core"),
  title = c("Exploratory Data Analysis",
  "Modern Applied Statistics ...",
  "LISP-STAT",
  "Spatial Statistics", "Stochastic Simulation",
  "Interactive Data Analysis",
  "An Introduction to R")
)
add_columns(books, authors)

# Just for fun. Examples borrowed from Microsoft Excel.
# It is not the R way of doing things.

# Example 2
ex2 = utils::read.table(header = TRUE, text = "
  Item_ID Item Cost Markup
ST-340 Stroller 145.67 0.30
BI-567 Bib 3.56 0.40
DI-328 Diapers 21.45 0.35
WI-989 Wipes 5.12 0.40
AS-469 Aspirator 2.56 0.45"
, stringsAsFactors = FALSE)
# Calculates the retail price of diapers by adding the markup percentage to the cost.
vlookup("DI-328", ex2, 3) * (1 + vlookup("DI-328", ex2, 4)) # 28.9575

# Calculates the sale price of wipes by subtracting a specified discount from
# the retail price.
(vlookup("WI-989", ex2, "Cost") * (1 + vlookup("WI-989", ex2, "Markup"))) * (1 - 0.2) # 5.7344

A2 = ex2[1, "Item_ID"]
A3 = ex2[2, "Item_ID"]

# If the cost of an item is greater than or equal to $20.00, displays the string
# "Markup is nn%"; otherwise, displays the string "Cost is under $20.00".
ifelse(vlookup(A2, ex2, "Cost") >= 20,
      paste0("Markup is ", 100 * vlookup(A2, ex2, "Markup"), ", ", "Cost is under ", "$20.00"),
      "Markup is under $20.00") # Markup is 30%

# If the cost of an item is greater than or equal to $20.00, displays the string
# "Markup is nn%"; otherwise, displays the string "Cost is $n.nn".
ifelse(vlookup(A3, ex2, "Cost") >= 20,
      paste0("Markup is: ", 100 * vlookup(A3, ex2, "Markup"), ", ", "Cost is $", vlookup(A3, ex2, "Cost")),
      "Cost is $n.nn") # Cost is $3.56

# Example 3
ex3 = utils::read.table(header = TRUE, text = 
  ID Last_name First_name Title Birth_date
  1 Davis Sara 'Sales Rep.' 12/8/1968
  2 Fontana Olivier 'V.P. of Sales' 2/19/1952
  3 Leal Karina 'Sales Rep.' 8/30/1963
  4 Patten Michael 'Sales Rep.' 9/19/1958
  5 Burke Brian 'Sales Mgr.' 3/4/1955
  6 Sousa Luis 'Sales Rep.' 7/2/1963
  ", stringsAsFactors = FALSE)

# If there is an employee with an ID of 5, displays the employee's last name;
# otherwise, displays the message "Employee not found".
if_na(vlookup(5, ex3, "Last_name"), "Employee not found") # Burke

# Many employees
if_na(vlookup(1:10, ex3, "Last_name"), "Employee not found")

# For the employee with an ID of 4, concatenates the values of three cells into
# a complete sentence.
paste0(vlookup(4, ex3, "First_name"), ", ",
       vlookup(4, ex3, "Last_name"), ", ",
       vlookup(4, ex3, "Title")) # Michael Patten is a Sales Rep.
**Description**

For the data frame `cond` will be evaluated in the data.frame’s context. So columns can be referred as variables in the expression (see the examples). If `data` is list then `where` will be applied to each element of the list. For other types (vector/matrix) there is no non-standard evaluation. There is a special constant `.N` which equals to number of rows in `data` for usage in `cond` expression. `.where` is version for working with default dataset. See `default_dataset`.

**Usage**

```r
where(data, cond)
.where(cond)
```

**Arguments**

- `data`: data.frame/matrix/vector/list to be subsetted
- `cond`: logical or numeric expression indicating elements or rows to keep: missing values (NA) are taken as FALSE. If `data` is data.frame then `cond` will be evaluated in the scope of the `data`.

**Value**

data.frame/matrix/vector/list which contains just selected rows.

**Examples**

```r
# leave only 'setosa'
where(iris, Species == "setosa")
# leave only first five rows
where(iris, 1:5)

# example of .N usage.
set.seed(42)
train = where(iris, sample(.N, 100))
str(train)

set.seed(42)
test = where(iris, -sample(.N, 100))
str(test)

# list example
set.seed(123)
rand_matr = matrix(sample(10, 60, replace = TRUE), ncol = 3)
rand_vec = sample(10, 20, replace = TRUE)
my_list = list(iris, rand_matr, rand_vec)
# two random elements from the each list item
where(my_list, sample(.N, 2))
```
**window_fun**

*Function over grouping variables (window function)*

**Description**

This is faster version of `ave`. `window_fun` applies function to every subset of `x` and return vector of the same length as `x`.

**Usage**

`window_fun(x, ...)`

**Arguments**

- `x` A vector
- `...` Grouping variables all of the same length as `x` or length 1 and function as last argument.

**Value**

vector of the same length as `x`

**Examples**

```r
window_fun(1:3, mean) # no grouping -> grand mean
attach(warpbreaks)
window_fun(breaks, wool, mean)
window_fun(breaks, tension, function(x) mean(x, trim = 0.1))
detach(warpbreaks)
```

----

**write_labelled_csv**

*Write labelled data to file or export file to SPSS syntax.*

**Description**

- `write_labelled_csv` and `read_labelled_csv` writes csv file with labels. By default labels are stored in the commented lines at the beginning of the file before the data part. *_csv2 write and read data with a semicolon separator and comma as decimal delimiter. *_tab/_tab2 write and read data with 'tab' separator and "."," as decimal delimiter.
- `write_labelled_xlsx` and `read_labelled_xlsx` write and read labelled 'xlsx' format. It is a simple Excel file with data and labels on separate sheets. It can help you with labelled data exchange in the corporate environment.
write_labelled_csv

- `write_labelled_fst` and `read_labelled_fst` write and read labelled data in the ‘fst’ format. See Fst Package. Data and labels are stored in the separate files. With ‘fst’ format you can read and write a huge amount of data very quickly.
- `write_labelled_spss` write ‘csv’ file with SPSS syntax for reading it. You can use it for the data exchange with SPSS.
- `create_dictionary` and `apply_dictionary` make data.frame with dictionary, e. g. variable and value labels for each variable. See format description in the ‘Details’ section.
- `write_labels` and `write_labels_spss` Write R code and SPSS syntax for labelling data. It allows to extract labels from *.sav files that come without accompanying syntax.
- `old_write_labelled_csv` and `old_read_labelled_csv` Read and write labelled ‘csv’ in format of the ‘expss’ version before 0.9.0.

Usage

```r
write_labelled_csv(
  x,
  filename,
  remove_new_lines = TRUE,
  single_file = TRUE,
  ...
)

write_labelled_csv2(
  x,
  filename,
  remove_new_lines = TRUE,
  single_file = TRUE,
  ...
)

write_labelled_tab(
  x,
  filename,
  remove_new_lines = TRUE,
  single_file = TRUE,
  ...
)

write_labelled_tab2(
  x,
  filename,
  remove_new_lines = TRUE,
  single_file = TRUE,
  ...
)

write_labelled_xlsx(
  x,
```
write_labelled_csv

    filename,
data_sheet = "data",
dict_sheet = "dictionary",
    remove_repeated = FALSE,
    use_references = TRUE
)

write_labelled_fst(x, filename, ...)

read_labelled_csv(filename, undouble_quotes = TRUE, ...)

read_labelled_csv2(filename, undouble_quotes = TRUE, ...)

read_labelled_tab(filename, undouble_quotes = TRUE, ...)

read_labelled_tab2(filename, undouble_quotes = TRUE, ...)

read_labelled_xlsx(filename, data_sheet = 1, dict_sheet = "dictionary")

read_labelled_fst(filename, ...)

write_labelled_spss(
    x,
    filename,
    fileEncoding = "",
    remove_new_lines = TRUE,
    ...
)

write_labels_spss(x, filename)

write_labels(x, filename, fileEncoding = "")

create_dictionary(x, remove_repeated = FALSE, use_references = TRUE)

apply_dictionary(x, dict)

old_write_labelled_csv(
    x,
    filename,
    fileEncoding = "",
    remove_new_lines = TRUE,
    ...
)

old_read_labelled_csv(filename, fileEncoding = "", undouble_quotes = TRUE, ...)
Arguments

x  data.frame to be written/data.frame whose labels to be written
filename the name of the file which the data are to be read from/write to.
remove_new_lines A logical indicating should we replace new lines with spaces in the character variables. TRUE by default.
single_file logical. TRUE by default. Should we write labels into the same file as data? If it is FALSE dictionary will be written in the separate file.
... additional arguments for fwrite/fread, e. g. column separator, decimal separator, encoding and etc.
data_sheet character "data" by default. Where data will be placed in the "*.xlsx" file.
dict_sheet character "dictionary" by default. Where dictionary will be placed in the "*.xlsx" file.
remove_repeated logical. FALSE by default. If TRUE then we remove repeated variable names. It makes a dictionary to look nicer for humans but less convenient for usage.
use_references logical. When TRUE (default) then if the variable has the same value labels as the previous variable, we use reference to this variable. It makes dictionary significantly more compact for datasets with many variables with the same value labels.
undouble_quotes A logical indicating should we undouble quotes which were escaped by doubling. TRUE by default. Argument will be removed when data.table issue #1109 will be fixed.
fileEncoding character string: if non-empty declares the encoding to be used on a file (not a connection) so the character data can be re-encoded as they are written. Used for writing dictionary. See file.
dict data.frame with labels - a result of create_dictionary.

Details

Dictionary is a data.frame with the following columns:

- variable variable name in the data set. It can be omitted (NA). In this case name from the previous row will be taken.
- value code for label in the column 'label'.
- label in most cases it is value label but its meaning can be changed by the column 'meta'.
- meta if it is NA then we have value label in the 'label' column. If it is 'var1ab', then there is a variable label in the 'label' column and column 'value' is ignored. If it is 'reference', then there is a variable name in the 'label' column and we use value labels from this variable, column 'value' is ignored.

Value

Functions for writing invisibly return NULL. Functions for reading return labelled data.frame.
Examples

```r
## Not run:
data(mtcars)
mtcars = modify(mtcars,
  var_lab(mpg) = "Miles/(US) gallon"
  var_lab(cyl) = "Number of cylinders"
  var_lab(disp) = "Displacement (cu.in.)"
  var_lab(hp) = "Gross horsepower"
  var_lab(drat) = "Rear axle ratio"
  var_lab(wt) = "Weight (lb/1000)"
  var_lab(qsec) = "1/4 mile time"
  var_lab(vs) = "Engine"
  val_lab(vs) = c("V-engine" = 0,
                   "Straight engine" = 1)
  var_lab(am) = "Transmission"
  val_lab(am) = c(automatic = 0,
                  manual=1)
  var_lab(gear) = "Number of forward gears"
  var_lab(carb) = "Number of carburetors"
})
write_labelled_csv(mtcars, "mtcars.csv")
new_mtcars = read_labelled_csv("mtcars.csv")
str(new_mtcars)
# identically, for xlsx
write_labelled_xlsx(mtcars, "mtcars.xlsx")
new_mtcars = read_labelled_xlsx("mtcars.xlsx")
str(new_mtcars)
# to SPSS syntax
write_labelled_spss(mtcars, "mtcars.csv")
```

## End(Not run)

---

### w_mean

**Compute various weighted statistics**

**Description**

- `w_mean` weighted mean of a numeric vector
- `w_sd` weighted sample standard deviation of a numeric vector
- `w_var` weighted sample variance of a numeric vector
- `w_se` weighted standard error of a numeric vector
- `w_median` weighted median of a numeric vector
- `w_mad` weighted mean absolute deviation from median of a numeric vector
• w_sum weighted sum of a numeric vector
• w_n weighted number of values of a numeric vector
• w_cov weighted covariance matrix of a numeric matrix/data.frame
• w_cor weighted Pearson correlation matrix of a numeric matrix/data.frame
• w_pearson shortcut for w_cor. Weighted Pearson correlation matrix of a numeric matrix/data.frame
• w_spearman weighted Spearman correlation matrix of a numeric matrix/data.frame

Usage

w_mean(x, weight = NULL, na.rm = TRUE)
w_median(x, weight = NULL, na.rm = TRUE)
w_var(x, weight = NULL, na.rm = TRUE)
w_sd(x, weight = NULL, na.rm = TRUE)
w_se(x, weight = NULL, na.rm = TRUE)
w_mad(x, weight = NULL, na.rm = TRUE)
w_sum(x, weight = NULL, na.rm = TRUE)
w_n(x, weight = NULL, na.rm = TRUE)
unweighted_valid_n(x, weight = NULL)
valid_n(x, weight = NULL)
w_max(x, weight = NULL, na.rm = TRUE)
w_min(x, weight = NULL, na.rm = TRUE)
w_cov(x, weight = NULL, use = c("pairwise.complete.obs", "complete.obs"))
w_cor(x, weight = NULL, use = c("pairwise.complete.obs", "complete.obs"))
w_pearson(x, weight = NULL, use = c("pairwise.complete.obs", "complete.obs"))
w_spearman(x, weight = NULL, use = c("pairwise.complete.obs", "complete.obs"))

Arguments

x a numeric vector (matrix/data.frame for correlations) containing the values whose weighted statistics is to be computed.
weight a vector of weights to use for each element of x. Cases with missing, zero or negative weights will be removed before calculations. If weight is missing then unweighted statistics will be computed.
na.rm: a logical value indicating whether NA values should be stripped before the computation proceeds. Note that contrary to base R statistic functions the default value is TRUE (remove missing values).

use: "pairwise.complete.obs" (default) or "complete.obs". In the first case the correlation or covariance between each pair of variables is computed using all complete pairs of observations on those variables. If use is "complete.obs" then missing values are handled by casewise deletion.

Details

If argument of correlation functions is data.frame with variable labels then variables names will be replaced with labels. If this is undesirable behavior use unvr function: w_cor(unvr(x)). Weighted spearman correlation coefficients are calculated with rounded to nearest integer weights. It gives the same result as in SPSS Statistics software. By now this algorithm is not memory efficient.

Value

a numeric value of length one/correlation matrix

Examples

data(mtcars)
dfs = mtcars %>% keep(mpg, disp, hp, wt)

with(dfs, w_mean(hp, weight = 1/wt))

# apply labels
dfs = modify(dfs, {
    var_lab(mpg) = "Miles/(US) gallon"
    var_lab(disp) = "Displacement (cu.in.)"
    var_lab(hp) = "Gross horsepower"
    var_lab(wt) = "Weight (1000 lbs)"
})

# weighted correlations with labels
w_cor(dfs, weight = 1/dfs$wt)

# without labels
w_cor(unvr(dfs), weight = 1/dfs$wt)

xl_write

Description

Note that openxlsx package is required for these functions. It can be install by printing install.packages('openxlsx') in the console. On Windows system you also may need to install rtools. You can export several tables at once by combining them in a list. See examples. If you need to write all tables to the single sheet you can use xl_write_file. It automatically creates workbook, worksheet and save *.xlsx file for you.
Usage

\texttt{xl\_write(obj, wb, sheet, row = 1, col = 1, ...)}

\texttt{xl\_write\_file(obj, filename, sheetname = "Tables", ...)}

## Default S3 method:
\texttt{xl\_write(}
  \texttt{obj,}
  \texttt{wb,}
  \texttt{sheet,}
  \texttt{row = 1,}
  \texttt{col = 1,}
  \texttt{rownames = FALSE,}
  \texttt{colnames = !is.atomic(obj),}
  \texttt{...}
\texttt{)}

## S3 method for class 'list'
\texttt{xl\_write(obj, wb, sheet, row = 1, col = 1, gap = 1, ...)}

## S3 method for class 'etable'
\texttt{xl\_write(}
  \texttt{obj,}
  \texttt{wb,}
  \texttt{sheet,}
  \texttt{row = 1,}
  \texttt{col = 1,}
  \texttt{remove\_repeated = c(\"all\", \"rows\", \"columns\", \"none\"),}
  \texttt{format\_table = TRUE,}
  \texttt{borders = list(border\_Colour = \"black\", border\_Style = \"thin\"),}
  \texttt{header\_format = openxlsx::createStyle(fg\_Fill = \#EBEBEB, halign = \"left\", wrap\_Text = FALSE),}
  \texttt{main\_format = openxlsx::createStyle(halign = \"right\", num\_Fmt = format(0, nsmall = get\_expss\_digits())),}
  \texttt{row\_labels\_format = openxlsx::createStyle(halign = \"left\"),}
  \texttt{total\_format = openxlsx::createStyle(fg\_Fill = \#EBEBEB, border = \"TopBottom\",}
  \texttt{border\_Style = \"thin\", halign = \"right\", num\_Fmt = \"0\"),}
  \texttt{total\_row\_labels\_format = openxlsx::createStyle(fg\_Fill = \#EBEBEB, border = \"TopBottom\",}
  \texttt{border\_Style = \"thin\", halign = \"left\"),}
  \texttt{top\_left\_corner\_format = header\_format,}
  \texttt{row\_symbols\_to\_remove = NULL,}
  \texttt{col\_symbols\_to\_remove = NULL,}
  \texttt{other\_rows\_formats = NULL,}
  \texttt{other\_row\_labels\_formats = NULL,}
  \texttt{other\_cols\_formats = NULL,}
  \texttt{other\_col\_labels\_formats = NULL,}
  \texttt{additional\_cells\_formats = NULL,}
  \texttt{...}
## S3 method for class 'with_caption'
xl_write(
  obj,
  wb,
  sheet,
  row = 1,
  col = 1,
  remove_repeated = c("all", "rows", "columns", "none"),
  format_table = TRUE,
  borders = list(borderColour = "black", borderStyle = "thin"),
  header_format = openxlsx::createStyle(fgFill = "#EBEBEB", halign = "left", wrapText = FALSE),
  main_format = openxlsx::createStyle(halign = "right", numFmt = format(0, nsmall = get_expss_digits())),
  row_labels_format = openxlsx::createStyle(halign = "left"),
  total_format = openxlsx::createStyle(fgFill = "#EBEBEB", border = "TopBottom", borderStyle = "thin", halign = "right", numFmt = "0"),
  total_row_labels_format = openxlsx::createStyle(fgFill = "#EBEBEB", border = "TopBottom", borderStyle = "thin", halign = "left"),
  top_left_corner_format = header_format,
  row_symbols_to_remove = NULL,
  col_symbols_to_remove = NULL,
  other_rows_formats = NULL,
  other_row_labels_formats = NULL,
  other_cols_formats = NULL,
  other_col_labels_formats = NULL,
  additional_cells_formats = NULL,
  caption_format = openxlsx::createStyle(textDecoration = "bold", halign = "left"),
  ...
)
remove_repeated
Should we remove duplicated row or column labels in the rows/columns of the
table? Possible values: "all", "rows", "columns", "none".

format_table
logical should we format table? If FALSE all format arguments will be ignored.

borders
list Style of the table borders. List with two named elements: borderColour
and borderStyle. For details see createStyle function. If it is NULL then no
table borders will be produced.

header_format
table header format - result of the createStyle function.

main_format
result of the createStyle function. Format of the table main area except total
rows. Total rows is rows which row labels contain '#'.

row_labels_format
result of the createStyle function. Format of the row labels area except total
rows. Total rows is rows which row labels contain '#'.

total_format
result of the createStyle function. Format of the total rows in the table main area.
Total rows is rows which row labels contain '#'.

total_row_labels_format
result of the createStyle function. Format of the total rows in the row labels area.
Total rows is rows which row labels contain '#'.

top_left_corner_format
result of the createStyle function.

row_symbols_to_remove
character vector. Perl-style regular expressions for substrings which will be re-
moved from row labels.

col_symbols_to_remove
character vector. Perl-style regular expressions for substrings which will be re-
moved from column names.

other_rows_formats
named list. Names of the list are perl-style regular expression patterns, items of
the list are results of the createStyle function. Rows in the main area which row
labels contain pattern will be formatted according to the appropriate style.

other_row_labels_formats
named list. Names of the list are perl-style regular expression patterns, items of
the list are results of the createStyle function. Rows in the row labels area which
row labels contain pattern will be formatted according to the appropriate style.

other_cols_formats
named list. Names of the list are perl-style regular expression patterns, items of
the list are results of the createStyle function. Columns in the main area which
column labels contain pattern will be formatted according to the appropriate
style.

other_col_labels_formats
named list. Names of the list are perl-style regular expression patterns, items of
the list are results of the createStyle function. Columns in the header area which
column labels contain pattern will be formatted according to the appropriate
style.
additional_cells_formats

- list Each item of the list is list which consists of two elements. First element is two columns matrix or data.frame with row number and column numbers in the main area of the table. Such matrix can be produced with code `which(logical_condition, arr.ind = TRUE)`. Instead of matrix one can use function which accepts original table (obj) and return such matrix. Second element is result of the `createStyle` function. Cells in the main area will be formatted according to this style.

caption_format result of the `createStyle` function.

Value

- invisibly return vector with rows and columns (c(rows, columns)) occupied by outputted object.

Examples

```r
## Not run:
library(openxlsx)
data(mtcars)
# add labels to dataset
mtcars = apply_labels(mtcars,
  mpg = "Miles/(US) gallon",
  cyl = "Number of cylinders",
  disp = "Displacement (cu.in.)",
  hp = "Gross horsepower",
  drat = "Rear axle ratio",
  wt = "Weight (lb/1000)",
  qsec = "1/4 mile time",
  vs = "Engine",
  vs = c("V-engine" = 0,
    "Straight engine" = 1),
  am = "Transmission",
  am = c("Automatic" = 0,
    "Manual"=1),
  gear = "Number of forward gears",
  carb = "Number of carburetors"
)

# create table with caption
mtcars_table = calc_cro_cpct(mtcars,
  cell_vars = list(cyl, gear),
  col_vars = list(total(), am, vs))

set_caption("Table 1")

wb = createWorkbook()
sh = addWorksheet(wb, "Tables")
# export table
xl_write(mtcars_table, wb, sh)
saveWorkbook(wb, "table1.xlsx", overwrite = TRUE)
```
## quick export
xl_write_file(mtcars_table, "table1.xlsx")

## custom cells formatting
wb = createWorkbook()
sh = addWorksheet(wb, "Tables")

# we want to mark cells which are greater than total column
my_formatter = function(tbl){
  greater_than_total = tbl[, -1] > tbl[[2]]
  which(greater_than_total, arr.ind = TRUE)
}

# export table
xl_write(mtcars_table, wb, sh,
  additional_cells_formats = list(
    list(my_formatter, createStyle(textDecoration = "bold", fontColour = "blue"))
  )
)

saveWorkbook(wb, "table_with_additional_format.xlsx", overwrite = TRUE)

## automated report generation on multiple variables with the same banner

banner = calc(mtcars, list(total(), am, vs))

# create list of tables
list_of_tables = lapply(mtcars, function(variable) {
  if(length(unique(variable))<7){
    cro_cpct(variable, banner) %>% significance_cpct()
  } else {
    # if number of unique values greater than seven we calculate mean
    cro_mean_sd_n(variable, banner) %>% significance_means()
  }
})

wb = createWorkbook()
sh = addWorksheet(wb, "Tables")
# export list of tables with additional formatting
xl_write(list_of_tables, wb, sh,
  # remove ' #' sign from totals
  col_symbols_to_remove = "#",
  row_symbols_to_remove = "#",
  # format total column as bold
  other_col_labels_formats = list("#" = createStyle(textDecoration = "bold")),
  other_cols_formats = list("#" = createStyle(textDecoration = "bold")),
)

saveWorkbook(wb, "report.xlsx", overwrite = TRUE)

## End(Not run)
Index

datasets
  if_na, 60
  product_test, 76
  recode, 80
  vars, 119
... , 16
.. (vars), 119
.. f, 67
.. f (vars), 119
.. p, 67
.. p (vars), 119
.. t, 67
.. t (vars), 119
  .add_columns (vlookup), 126
  .add_rows (add_rows), 3
  .apply_labels (apply_labels), 4
  .calc (experimental), 46
  .calculate (experimental), 46
  .compute, 42
  .compute (experimental), 46
  .cro (experimental), 46
  .cro_cases (experimental), 46
  .cro_cpct (experimental), 46
  .cro_fun (experimental), 46
  .cro_fun_df (experimental), 46
  .cro_mean (experimental), 46
  .cro_median_sd_n (experimental), 46
  .cro_median (experimental), 46
  .cro_rpct (experimental), 46
  .cro_sum (experimental), 46
  .cro_tpct (experimental), 46
  .do_if (experimental), 46
  .do_repeat (do_repeat), 43
  .except (keep), 63
  .fre (experimental), 46
  .if_val (experimental), 46
  .keep (keep), 63
  .modify (experimental), 46
  .modify_if (experimental), 46
  .recode (experimental), 46
  .sheet (sheet), 87
  .sort_asc (sort_asc), 88
  .sort_desc (sort_asc), 88
  .val_lab (experimental), 46
  .var_lab (experimental), 46
  .where (where), 130
% (vectors), 124
%add_rows% (add_rows), 3
%calc% (compute), 16
%calculate% (compute), 16
%col_in% (count_if), 19
%d% (vectors), 124
%d% (vectors), 124
%has% (count_if), 19
%e% (vectors), 124
%if_na% (if_na), 60
%into% (recode), 80
%merge% (merge.etable), 66
%mis_val% (if_na), 60
%n_d% (vectors), 124
%nest% (nest), 70
%r% (vectors), 124
%row_in% (count_if), 19
%thru% (criteria), 25
%to% (vars), 119
%to_list% (vars), 119
%u% (vectors), 124
%use_labels% (compute), 16
%v% 25, 28
%to%, 43, 94

add_columns (vlookup), 126
add_labelled_class (var_lab), 122
add_rows, 3, 30
add_sig_labels
  (tab_significance_options), 103
add_val_lab (val_lab), 117
add_val_lab <- (val_lab, 117
all, 92
all_in_col (sum_row), 92
all_in_row (sum_row), 92
and (criteria), 25
any, 92
any_in_col (sum_row), 92
any_in_row, 50
any_in_row (sum_row), 92
append, 124
apply_col (sum_row), 92
apply_col_if (count_if), 19
apply_dictionary (write_labelled_csv), 131
apply_labels, 4, 117, 123
apply_row (sum_row), 92
apply_row_if (count_if), 19
as.category, 5, 10, 67, 68
as.criterion (criteria), 25
as.data.frame, 87
as.datatable_widget, 7, 51, 86
as.dichotomy, 6, 8, 67, 68
as.etable, 11
as.factor, 51, 53
as.labelled, 12
as.ordered, 53
as.sheet (sheet), 87
as_hux.etable (as_huxtable.etable), 13
as_huxtable.etable, 13
as_is (do_repeat), 43
assign, 120
autonum (val_lab), 117
ave, 131
c, 36
calc (compute), 16
calc_cro (cro), 29
calc_cro_cases (cro), 29
calc_cro_cpct (cro), 29
calc_cro_cpct_responses (cro), 29
calc_cro_fun (cro_fun), 35
calc_cro_fun_df (cro_fun), 35
calc_cro_mean (cro_fun), 35
calc_cro_mean_sd_n (cro_fun), 35
calc_cro_median (cro_fun), 35
calc_cro_pearson (cro_fun), 35
calc_cro_rptct (cro), 29
calc_cro_spearman (cro_fun), 35
calc_cro_sum (cro_fun), 35
calc_cro_tpct (cro), 29
calculate, 48, 89, 120, 121
calculate (compute), 16
cell_chisq (tab_significance_options), 103
chisq.test, 103, 108, 110
combine_functions, 98
combine_functions (cro_fun), 35
calculate_means, 110
calculate_means (compare_proportions), 14
calculate_proportions, 14, 110
complete_cases, 60
compute, 16, 44, 50, 89, 92, 94, 121
count (criteria), 25
copy (recode), 80
count_col_if (count_if), 19
count_if, 19, 25, 28, 50, 94
count_row_if, 50
count_row_if (count_if), 19
creates_dictionary (write_labelled_csv), 131
creates_style, 140, 141
createsWorkbook, 139
criteria, 21, 25, 43, 60, 63, 64, 80, 82, 88,
112, 119, 120, 125
cro, 3, 4, 7, 29, 40, 45, 48, 50, 56, 58, 59, 66,
67, 70, 71, 90, 99, 112, 139
cro_cases, 103, 109, 110
cro_cases (cro), 29
cro_cpct, 103, 109, 110
cro_cpct (cro), 29
cro_cpct_responses (cro), 29
cro_fun, 3, 4, 34, 35, 45, 48, 50, 66, 67, 90,
98, 99
cro_fun_df, 45
cro_fun_df (cro_fun), 35
cro_mean (cro_fun), 35
cro_mean_sd_n, 103, 109, 110
cro_mean_sd_n (cro_fun), 35
cro_median (cro_fun), 35
cro_ppearson (cro_fun), 35
cro_rptct (cro), 29
cro_spearman (cro_fun), 35
cro_sum (cro_fun), 35
cro_tptct (cro), 29
data.frame, 87
data.table, 39, 98
datatable, 7
is_max (criteria), 25
is_min (criteria), 25
is_na (criteria), 25
items (criteria), 25

keep, 63, 89, 120, 121
knit_print.etable (htmlTable.etable), 56
knit_print.with_caption
(htmlTable.etable), 56

lab_num (val_lab), 117
le (criteria), 25
less (criteria), 25
less_or_equal (criteria), 25
like (criteria), 25
lo (recode), 80
lt (criteria), 25
lte (criteria), 25

make_labels (val_lab), 117
make_subheadings (split_labels), 90
match_col (match_row), 64
match_row, 25, 28, 50, 64
max_col (sum_row), 92
max_col_if (count_if), 19
max_if, 94
max_if (count_if), 19
max_row, 50
max_row (sum_row), 92
max_row_if, 50
max_row_if (count_if), 19
mdset, 6, 10, 30, 33, 39, 97, 98
mdset (mrset), 67
mdset_f (mrset), 67
mdset_t (mrset), 67
mean_col (sum_row), 92
mean_col_if (count_if), 19
mean_if, 94
mean_if (count_if), 19
mean_row, 50
mean_row (sum_row), 92
mean_row_if, 50
mean_row_if (count_if), 19
median_col (sum_row), 92
median_col_if (count_if), 19
median_if, 94
median_if (count_if), 19
median_row (sum_row), 92
median_row_if (count_if), 19
merge, 66
merge.etable, 30, 66
min_col (sum_row), 92
min_col_if (count_if), 19
min_if, 94
min_if (count_if), 19
min_row, 50
min_row (sum_row), 92
min_row_if, 50
min_row_if (count_if), 19
mis_val (if_na), 60
mis_val< (if_na), 60
modify, 48, 120
modify (compute), 16
modify_if, 48, 120
modify_if (compute), 16
mrset, 6, 10, 30, 33, 39, 97, 98, 104
mrset_f (mrset), 67
mrset_p (mrset), 67
mrset_t (mrset), 67

n2l (names2labels), 68
n_diff (vectors), 124
n_intersect (vectors), 124
na_if, 25, 28, 98
na_if (if_na), 60
na_if< (if_na), 60
name_dots, 69
names2labels, 53, 68, 116
ne (criteria), 25
neq (criteria), 25
nest, 30, 70
net, 71
not (criteria), 25
not_equals (criteria), 25
not_na (criteria), 25
num_lab (val_lab), 117

old_read_labelled_csv
(write_labelled_csv), 131
old_write_labelled_csv
(write_labelled_csv), 131
or (criteria), 25
ordered, 53
other (criteria), 25

perl (criteria), 25
prepend_all (prepend_values), 75
prepend_names (prepend_values), 75
prepend_values, 75
product_test, 76
prop, 77
prop.test, 14, 15, 103, 110
prop_col (prop), 77
prop_row (prop), 77
q.c, 78
q.e (q.c), 78
rbind, 3, 4
read.spss, 80
read.table, 113, 114
read_labelled_csv (write_labelled_csv), 131
read_labelled_csv2
  (write_labelled_csv), 131
read_labelled_fst (write_labelled_csv), 131
read_labelled_tab (write_labelled_csv), 131
read_labelled_tab2
  (write_labelled_csv), 131
read_labelled.xlsx
  (write_labelled_csv), 131
read_spss, 79
read.spss.to.list (read_spss), 79
rec (recode), 80
rec<- (recode), 80
recode, 25, 28, 48, 50, 60, 80
recode<- (recode), 80
ref, 42, 84
ref<- (ref), 84
regex (criteria), 25
rep, 125
repr_html.etable (htmlTable.etable), 56
repr_html.with_caption
  (htmlTable.etable), 56
repr.text.etable (htmlTable.etable), 56
repr.text.with_caption
  (htmlTable.etable), 56
sd_col (sum_row), 92
sd_col_if (count_if), 19
sd_if, 94
sd_if (count_if), 19
sd_row (sum_row), 92
sd_row_if (count_if), 19
set_caption, 86
set_default_properties, 52
set_val_lab (val_lab), 117
set_var_lab (var_lab), 122
sheet, 87
significance, 94, 99
significance
  (tab_significance_options), 103
significance_cases
  (tab_significance_options), 103
significance_cell_chisq
  (tab_significance_options), 103
significance_cpct, 14, 15
significance_cpct
  (tab_significance_options), 103
significance_means, 14, 15, 35
significance_means
  (tab_significance_options), 103
sort_asc, 88
sort_desc (sort_asc), 88
split, 89
split_by, 89
split_columns (split_labels), 90
split_labels, 90
split_off (split_by), 89
split_table_to_df (split_labels), 90
strsplit, 91
subset, 48
subtotal (net), 71
sum_col (sum_row), 92
sum_col_if (count_if), 19
sum_if, 94
sum_if (count_if), 19
sum_row, 50, 92
sum_row_if, 50
sum_row_if (count_if), 19
t.test, 14, 15, 103, 110
tab_caption (tables), 94
tab_cells (tables), 94
tab_cols (tables), 94
tab_last_add_sig_labels
  (tab_significance_options), 103
tab_last_hstack (tables), 94
tab_last_round
  (tab_significance_options), 103
tab_last_sig_cases
  (tab_significance_options), 103
<table>
<thead>
<tr>
<th>Function</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>tab_last_sig_cell_chisq</code></td>
<td>(tab_significance_options), 103</td>
</tr>
<tr>
<td><code>tab_last_sig_cpct</code></td>
<td>(tab_significance_options), 103</td>
</tr>
<tr>
<td><code>tab_last_sig_means</code></td>
<td>(tab_significance_options), 103</td>
</tr>
<tr>
<td><code>tab_last_vstack</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_mis_val</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_net_cells</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_net_cols</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_net_rows</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_pivot</code></td>
<td>109</td>
</tr>
<tr>
<td><code>tab_prepend_all</code></td>
<td>(prepend_values), 75</td>
</tr>
<tr>
<td><code>tab_prepend_names</code></td>
<td>(prepend_values), 75</td>
</tr>
<tr>
<td><code>tab_prepend_values</code></td>
<td>(prepend_values), 75</td>
</tr>
<tr>
<td><code>tab_row_label</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_rows</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_sort_asc</code></td>
<td>criteria</td>
</tr>
<tr>
<td><code>tab_sort_desc</code></td>
<td></td>
</tr>
<tr>
<td><code>tab_stat_cases</code></td>
<td>104</td>
</tr>
<tr>
<td><code>tab_stat_cpct</code></td>
<td>104</td>
</tr>
<tr>
<td><code>tab_stat_cpct_responses</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_fun</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_fun_df</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_max</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_median</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_min</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_rptct</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_sd</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_se</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_sum</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_tpct</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_unweighted_valid_n</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_stat_valid_n</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_subgroup</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_subtotal_cells</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_subtotal_cols</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_subtotal_rows</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>tab_total_label</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_total_row_position</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_total_statistic</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_transpose</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tab_weight</code></td>
<td>(tables), 94</td>
</tr>
<tr>
<td><code>tables</code></td>
<td>3, 4, 7, 30, 34, 40, 45, 50, 56, 67, 71, 73, 75, 94, 104, 110, 112</td>
</tr>
<tr>
<td><code>text_expand</code></td>
<td>67, 120</td>
</tr>
<tr>
<td><code>text_to_columns</code></td>
<td>113</td>
</tr>
<tr>
<td><code>text_to_columns_csv</code></td>
<td>(text_to_columns), 113</td>
</tr>
<tr>
<td><code>text_to_columns_csv2</code></td>
<td>(text_to_columns), 113</td>
</tr>
<tr>
<td><code>text_to_columns_tab</code></td>
<td>(text_to_columns), 113</td>
</tr>
<tr>
<td><code>text_to_columns_tab2</code></td>
<td>(text_to_columns), 113</td>
</tr>
<tr>
<td><code>thru</code></td>
<td>(criteria), 25</td>
</tr>
<tr>
<td><code>to</code></td>
<td>(criteria), 25</td>
</tr>
<tr>
<td><code>total</code></td>
<td>(cro), 29</td>
</tr>
<tr>
<td><code>unhide</code></td>
<td>(net), 71</td>
</tr>
<tr>
<td><code>unlab</code></td>
<td>115</td>
</tr>
<tr>
<td><code>unvl</code></td>
<td>115</td>
</tr>
<tr>
<td><code>unvr</code></td>
<td>(val_lab), 117</td>
</tr>
<tr>
<td><code>unvr</code></td>
<td>(val_lab), 122</td>
</tr>
<tr>
<td><code>unweighted_valid_n</code></td>
<td>(w_mean), 135</td>
</tr>
<tr>
<td><code>use_labels</code></td>
<td>(compute), 16</td>
</tr>
<tr>
<td><code>v2l</code></td>
<td>(values2labels), 116</td>
</tr>
<tr>
<td><code>v_diff</code></td>
<td>(vectors), 124</td>
</tr>
<tr>
<td><code>v_intersect</code></td>
<td>(vectors), 124</td>
</tr>
<tr>
<td><code>v_union</code></td>
<td>(vectors), 124</td>
</tr>
<tr>
<td><code>v_xor</code></td>
<td>(vectors), 124</td>
</tr>
<tr>
<td><code>val_lab</code></td>
<td>4, 48, 50, 53, 69, 80, 116, 117, 123</td>
</tr>
<tr>
<td><code>val_lab&lt;-</code></td>
<td>(val_lab), 117</td>
</tr>
<tr>
<td><code>valid</code></td>
<td>(if_na), 60</td>
</tr>
<tr>
<td><code>valid_n</code></td>
<td>(w_mean), 135</td>
</tr>
<tr>
<td><code>value_col_if</code></td>
<td>(match_row), 64</td>
</tr>
<tr>
<td><code>value_row_if</code></td>
<td>(match_row), 64</td>
</tr>
<tr>
<td><code>values2labels</code></td>
<td>53, 69, 116</td>
</tr>
<tr>
<td><code>var_lab</code></td>
<td>4, 48, 50, 53, 69, 80, 116, 117, 122</td>
</tr>
<tr>
<td><code>var_lab&lt;-</code></td>
<td>(var_lab), 122</td>
</tr>
<tr>
<td><code>vars</code></td>
<td>119</td>
</tr>
<tr>
<td><code>vars_list</code></td>
<td>(vars), 119</td>
</tr>
<tr>
<td><code>vectors</code></td>
<td>124</td>
</tr>
<tr>
<td><code>vlookup</code></td>
<td>50, 126</td>
</tr>
<tr>
<td><code>vlookup_df</code></td>
<td>(vlookup), 126</td>
</tr>
<tr>
<td><code>w_cor</code></td>
<td>(w_mean), 135</td>
</tr>
</tbody>
</table>
w_cov (w_mean), 135
w_mad (w_mean), 135
w_max (w_mean), 135
w_mean, 35, 135
w_median (w_mean), 135
w_min (w_mean), 135
w_n (w_mean), 135
w_pearson (w_mean), 135
w_sd (w_mean), 135
w_se (w_mean), 135
w_spearman (w_mean), 135
w_sum (w_mean), 135
w_var (w_mean), 135
when (criteria), 25
where, 48, 89, 120, 121, 129
window_fun, 131
write_labelled_csv, 131
write_labelled_csv2
   (write_labelled_csv), 131
write_labelled_fst
   (write_labelled_csv), 131
write_labelled_spss
   (write_labelled_csv), 131
write_labelled_tab
   (write_labelled_csv), 131
write_labelled_tab2
   (write_labelled_csv), 131
write_labelled_xlsx
   (write_labelled_csv), 131
write_labels (write_labelled_csv), 131
write_labels_spss (write_labelled_csv), 131
xl_write, 86, 137
xl_write_file (xl_write), 137