Package 'VIM'

July 21, 2025

```
Version 6.2.2
```

Title Visualization and Imputation of Missing Values

Depends R (>= 3.5.0),colorspace,grid

Imports car, grDevices, magrittr, robustbase, stats, sp, vcd,MASS,nnet,e1071,methods,Rcpp,utils,graphics,laeken,ranger, data.table(>= 1.9.4)

Suggests dplyr, tinytest, knitr, rmarkdown, reactable, covr, withr

Description New tools for the visualization of missing and/or imputed values are introduced, which can be used for exploring the data and the structure of the missing and/or imputed values. Depending on this structure of the missing values, the corresponding methods may help to identify the mechanism generating the missing values and allows to explore the data including missing values. In addition, the quality of imputation can be visually explored using various univariate, bivariate, multiple and multivariate plot methods. A graphical user interface available in the separate package VIMGUI allows an easy handling of the implemented plot methods.

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Description

This package introduces new tools for the visualization of missing or imputed values in , which can be used for exploring the data and the structure of the missing or imputed values. Depending on this structure, they may help to identify the mechanism generating the missing values or errors, which may have happened in the imputation process. This knowledge is necessary for selecting an appropriate imputation method in order to reliably estimate the missing values. Thus the visualization tools should be applied before imputation and the diagnostic tools afterwards.

Details

Detecting missing values mechanisms is usually done by statistical tests or models. Visualization of missing and imputed values can support the test decision, but also reveals more details about the data structure. Most notably, statistical requirements for a test can be checked graphically, and problems like outliers or skewed data distributions can be discovered. Furthermore, the included plot methods may also be able to detect missing values mechanisms in the first place.

A graphical user interface available in the package VIMGUI allows an easy handling of the plot methods. In addition, VIM can be used for data from essentially any field.

Package: VIM Version: 3.0.3 Date: 2013-01-09

Depends: R (>= 2.10),e1071,car, colorspace, nnet, robustbase, tcltk, tkrplot, sp, vcd, Rcpp

Imports: car, colorspace, grDevices, robustbase, stats, tcltk, sp, utils, vcd

License: GPL (>= 2)

URL: http://cran.r-project.org/package=VIM

Author(s)

Matthias Templ, Andreas Alfons, Alexander Kowarik, Bernd Prantner

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References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

M. Templ, A. Kowarik, P. Filzmoser (2011) Iterative stepwise regression imputation using standard and robust methods. *Journal of Computational Statistics and Data Analysis*, Vol. 55, pp. 2793-2806.

aggr

Aggregations for missing/imputed values

Description

Calculate or plot the amount of missing/imputed values in each variable and the amount of missing/imputed values in certain combinations of variables.

Print method for objects of class "aggr".

Summary method for objects of class "aggr".

Print method for objects of class "summary.aggr".

Usage

```
aggr(x, delimiter = NULL, plot = TRUE, ...)
## S3 method for class 'aggr'
plot(
  Х,
  col = c("skyblue", "red", "orange"),
  bars = TRUE,
  numbers = FALSE,
  prop = TRUE,
  combined = FALSE,
  varheight = FALSE,
  only.miss = FALSE,
  border = par("fg"),
  sortVars = FALSE,
  sortCombs = TRUE,
  ylabs = NULL,
  axes = TRUE,
```

```
labels = axes,
  cex.lab = 1.2,
  cex.axis = par("cex"),
  cex.numbers = par("cex"),
  gap = 4,
  ...
)

## S3 method for class 'aggr'
print(x, ..., digits = NULL)

## S3 method for class 'aggr'
summary(object, ...)

## S3 method for class 'summary.aggr'
print(x, ...)
```

Arguments

x an object of class "summary.aggr".

delimiter a character-vector to distinguish between variables and imputation-indices for

imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

plot a logical indicating whether the results should be plotted (the default is TRUE).

... Further arguments, currently ignored.

col a vector of length three giving the colors to be used for observed, missing and

imputed data. If only one color is supplied, it is used for missing and imputed data and observed data is transparent. If only two colors are supplied, the first one is used for observed data and the second color is used for missing and im-

puted data.

bars a logical indicating whether a small barplot for the frequencies of the different

combinations should be drawn.

numbers a logical indicating whether the proportion or frequencies of the different com-

binations should be represented by numbers.

prop a logical indicating whether the proportion of missing/imputed values and com-

binations should be used rather than the total amount.

combined a logical indicating whether the two plots should be combined. If FALSE, a

separate barplot on the left hand side shows the amount of missing/imputed values in each variable. If TRUE, a small version of this barplot is drawn on top of the plot for the combinations of missing/imputed and non-missing values.

See "Details" for more information.

varheight a logical indicating whether the cell heights are given by the frequencies of

occurrence of the corresponding combinations.

only.miss	a logical indicating whether the small barplot for the frequencies of the combinations should only be drawn for combinations including missing/imputed values (if bars is TRUE). This is useful if most observations are complete, in which case the corresponding bar would dominate the barplot such that the remaining bars are too compressed. The proportion or frequency of complete observations (as determined by prop) is then represented by a number instead of a bar.
border	the color to be used for the border of the bars and rectangles. Use border=NA to omit borders.
sortVars	a logical indicating whether the variables should be sorted by the number of missing/imputed values.
sortCombs	a logical indicating whether the combinations should be sorted by the frequency of occurrence.
ylabs	if combined is TRUE, a character string giving the y-axis label of the combined plot, otherwise a character vector of length two giving the y-axis labels for the two plots.
axes	a logical indicating whether axes should be drawn.
labels	either a logical indicating whether labels should be plotted on the x-axis, or a character vector giving the labels.
cex.lab	the character expansion factor to be used for the axis labels.
cex.axis	the character expansion factor to be used for the axis annotation.
cex.numbers	the character expansion factor to be used for the proportion or frequencies of the different combinations
gap	if combined is FALSE, a numeric value giving the distance between the two plots in margin lines.
digits	the minimum number of significant digits to be used (see print.default()).
object	an object of class "aggr".

Details

Often it is of interest how many missing/imputed values are contained in each variable. Even more interesting, there may be certain combinations of variables with a high number of missing/imputed values.

If combined is FALSE, two separate plots are drawn for the missing/imputed values in each variable and the combinations of missing/imputed and non-missing values. The barplot on the left hand side shows the amount of missing/imputed values in each variable. In the *aggregation plot* on the right hand side, all existing combinations of missing/imputed and non-missing values in the observations are visualized. Available, missing and imputed data are color coded as given by col. Additionally, there are two possibilities to represent the frequencies of occurrence of the different combinations. The first option is to visualize the proportions or frequencies by a small bar plot and/or numbers. The second option is to let the cell heights be given by the frequencies of the corresponding combinations. Furthermore, variables may be sorted by the number of missing/imputed values and combinations by the frequency of occurrence to give more power to finding the structure of missing/imputed values.

If combined is TRUE, a small version of the barplot showing the amount of missing/imputed values in each variable is drawn on top of the aggregation plot.

The graphical parameter oma will be set unless supplied as an argument.

Value

for aggr, a list of class "aggr" containing the following components:

- x the data used.
- combinations a character vector representing the combinations of variables.
- count the frequencies of these combinations.
- percent the percentage of these combinations.
- missings a data. frame containing the amount of missing/imputed values in each variable.
- tabcomb the indicator matrix for the combinations of variables.

a list of class "summary.aggr" containing the following components:

- missings a data. frame containing the amount of missing or imputed values in each variable.
- combinations a data.frame containing a character vector representing the combinations of variables along with their frequencies and percentages.

Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments labs and names.arg can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use ylabs and labels instead.

Author(s)

Andreas Alfons, Matthias Templ, modifications for displaying imputed values by Bernd Prantner Matthias Templ, modifications by Andreas Alfons and Bernd Prantner

Matthias Templ, modifications by Andreas Alfons

Andreas Alfons, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
print.aggr(), summary.aggr()
aggr()
print.summary.aggr(), aggr()
summary.aggr(), aggr()
Other plotting functions: barMiss(), histMiss(), marginmatrix(), marginplot(), matrixplot(),
mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

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Examples

```
data(sleep, package="VIM")
## for missing values
a <- aggr(sleep)
a
summary(a)

## for imputed values
sleep_IMPUTED <- kNN(sleep)
a <- aggr(sleep_IMPUTED, delimiter="_imp")
a
summary(a)

data(sleep, package = "VIM")
a <- aggr(sleep, plot=FALSE)
a

data(sleep, package = "VIM")
summary(aggr(sleep, plot=FALSE))

data(sleep, package = "VIM")
s <- summary(aggr(sleep, plot=FALSE))
s</pre>
```

alphablend

Alphablending for colors

Description

Convert colors to semitransparent colors.

Usage

```
alphablend(col, alpha = NULL, bg = NULL)
```

Arguments

col a vector specifying colors.

alpha a numeric vector containing the alpha values (between 0 and 1).

bg the background color to be used for alphablending. This can be used as a

workaround for graphics devices that do not support semitransparent colors.

Value

a vector containing the semitransparent colors.

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Author(s)

Andreas Alfons

Examples

```
alphablend("red", 0.6)
```

Animals_na

Animals_na

Description

Average log brain and log body weights for 28 Species

Format

A data frame with 28 observations on the following 2 variables.

lbody log body weight

Ibrain log brain weight

Details

The original data can be found in package MASS. 10 values on brain weight are set to be missing.

Source

P. J. Rousseeuw and A. M. Leroy (1987) Robust Regression and Outlier Detection. Wiley, p. 57.

References

Venables, W. N. and Ripley, B. D. (1999) Modern Applied Statistics with S-PLUS. Third Edition. Springer.

Templ, M. (2022) Visualization and Imputation of Missing Values. Springer Publishing. Upcoming book.

Examples

```
data(Animals_na)
aggr(Animals_na)
```

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barMiss

Barplot with information about missing/imputed values

Description

Barplot with highlighting of missing/imputed values in other variables by splitting each bar into two parts. Additionally, information about missing/imputed values in the variable of interest is shown on the right hand side.

Usage

```
barMiss(
  Χ,
  delimiter = NULL,
  pos = 1,
  selection = c("any", "all"),
  col = c("skyblue", "red", "skyblue4", "red4", "orange", "orange4"),
 border = NULL,
 main = NULL,
  sub = NULL,
 xlab = NULL
 ylab = NULL,
  axes = TRUE,
  labels = axes,
 only.miss = TRUE,
 miss.labels = axes,
  interactive = TRUE,
)
```

Arguments

x a vector, matrix or data. frame.

delimiter a character-vector to distinguish between variables and imputation-indices for

imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

pos a numeric value giving the index of the variable of interest. Additional variables

in x are used for highlighting.

selection the selection method for highlighting missing/imputed values in multiple additional variables. Possible values are "any" (highlighting of missing/imputed val-

ues in *any* of the additional variables) and "all" (highlighting of missing/imputed

values in *all* of the additional variables).

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col a vector of length six giving the colors to be used. If only one color is supplied, the bars are transparent and the supplied color is used for highlighting missing/imputed values. Else if two colors are supplied, they are recycled. border the color to be used for the border of the bars. Use border=NA to omit borders. main, sub main and sub title. axis labels. xlab, ylab axes a logical indicating whether axes should be drawn on the plot. labels either a logical indicating whether labels should be plotted below each bar, or a character vector giving the labels. only.miss logical; if TRUE, the missing/imputed values in the variable of interest are visualized by a single bar. Otherwise, a small barplot is drawn on the right hand side (see 'Details'). miss.labels either a logical indicating whether label(s) should be plotted below the bar(s) on the right hand side, or a character string or vector giving the label(s) (see 'Details'). a logical indicating whether variables can be switched interactively (see 'Deinteractive tails').

further graphical parameters to be passed to graphics::title() and graphics::axis().

Details

If more than one variable is supplied, the bars for the variable of interest are split according to missingness/number of imputed missings in the additional variables.

If only.miss=TRUE, the missing/imputed values in the variable of interest are visualized by one bar on the right hand side. If additional variables are supplied, this bar is again split into two parts according to missingness/number of imputed missings in the additional variables.

Otherwise, a small barplot consisting of two bars is drawn on the right hand side. The first bar corresponds to observed values in the variable of interest and the second bar to missing/imputed values. Since these two bars are not on the same scale as the main barplot, a second y-axis is plotted on the right (if axes=TRUE). Each of the two bars are again split into two parts according to missingness/number of imputed missings in the additional variables. Note that this display does not make sense if only one variable is supplied, therefore only.miss is ignored in that case.

If interactive=TRUE, clicking in the left margin of the plot results in switching to the previous variable and clicking in the right margin results in switching to the next variable. Clicking anywhere else on the graphics device quits the interactive session. When switching to a continuous variable, a histogram is plotted rather than a barplot.

Value

a numeric vector giving the coordinates of the midpoints of the bars.

Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the

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arguments axisnames, names.arg and names.miss can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use labels and miss.labels instead.

Author(s)

Andreas Alfons, modifications to show imputed values by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
spineMiss(), histMiss()
Other plotting functions: aggr(), histMiss(), marginmatrix(), marginplot(), matrixplot(),
mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

```
data(sleep, package = "VIM")
## for missing values
x <- sleep[, c("Exp", "Sleep")]
barMiss(x)
barMiss(x, only.miss = FALSE)

## for imputed values
x_IMPUTED <- kNN(sleep[, c("Exp", "Sleep")])
barMiss(x_IMPUTED, delimiter = "_imp")
barMiss(x_IMPUTED, delimiter = "_imp", only.miss = FALSE)</pre>
```

bcancer

Breast cancer Wisconsin data set

Description

Dataset containing the original Wisconsin breast cancer data.

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Format

A data frame with 699 observations on the following 11 variables.

```
clump_thickness as integer from 1 - 10
uniformity_cellsize as integer from 1 - 10
uniformity_cellshape as integer from 1 - 10
adhesion as integer from 1 - 10
epithelial_cellsize as integer from 1 - 10
bare_nuclei as integer from 1 - 10, includes 16 missings chromatin as integer from 1 - 10
normal_nucleoli as integer from 1 - 10
mitoses as integer from 1 - 10
class benign or malignant
```

References

The data downloaded and conditioned for R from the UCI machine learning repository, see https://archive.ics.uci.edu/ml/datas. This breast cancer databases was obtained from the University of Wisconsin Hospitals, Madison from Dr. William H. Wolberg. If you publish results when using this database, then please include this information in your acknowledgements. Also, please cite one or more of: O. L. Mangasarian and W. H. Wolberg: "Cancer diagnosis via linear programming", SIAM News, Volume 23, Number 5, September 1990, pp 1 & 18. William H. Wolberg and O.L. Mangasarian: "Multisurface method of pattern separation for medical diagnosis applied to breast cytology", Proceedings of the National Academy of Sciences, U.S.A., Volume 87, December 1990, pp 9193-9196. O. L. Mangasarian, R. Setiono, and W.H. Wolberg: "Pattern recognition via linear programming: Theory and application to medical diagnosis", in: "Large-scale numerical optimization", Thomas F. Coleman and Yuying Li, editors, SIAM Publications, Philadelphia 1990, pp 22-30. K. P. Bennett & O. L. Mangasarian: "Robust linear programming discrimination of two linearly inseparable sets", Optimization Methods and Software 1, 1992, 23-34 (Gordon & Breach Science Publishers).

Examples

data(bcancer)
aggr(bcancer)

bgmap

Backgound map

Description

Plot a background map.

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Usage

```
bgmap(map, add = FALSE, ...)
```

Arguments

map either a matrix or data. frame with two columns, a list with components x and y, or an object of any class that can be used for maps and provides its own plot method (e.g., "SpatialPolygons" from package sp). A list of the previously mentioned types can also be provided. add a logical indicating whether map should be added to an already existing plot (the

default is FALSE).

further arguments and graphical parameters to be passed to plot and/or graphics::lines(). . . .

Author(s)

Andreas Alfons

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. Journal of Advances in Data Analysis and Classification, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
growdotMiss(), mapMiss()
```

Examples

```
data(kola.background, package = "VIM")
bgmap(kola.background)
```

brittleness

Brittleness index data set

Description

A plastic product is produced in three parallel reactors (TK104, TK105, or TK107). For each row in the dataset, we have the same batch of raw material that was split, and fed to the 3 reactors. These values are the brittleness index for the product produced in the reactor. A simulated data set.

Format

A data frame with 23 observations on the following 3 variables.

TK104 Brittleness for batches of raw material in reactor 104

TK105 Brittleness for batches of raw material in reactor 105

TK107 Brittleness for batches of raw material in reactor 107

Source

https://openmv.net/info/brittleness-index

Examples

data(brittleness)
aggr(brittleness)

chorizonDL

C-horizon of the Kola data with missing values

Description

This data set is the same as in package mvoutlier, except that values below the detection limit are coded as NA.

Format

A data frame with 606 observations on the following 110 variables.

*ID a numeric vector

XCOO a numeric vector

YCOO a numeric vector

Ag a numeric vector

Ag_INAA a numeric vector

Al a numeric vector

Al2O3 a numeric vector

As a numeric vector

As_INAA a numeric vector

Au_INAA a numeric vector

B a numeric vector

Ba a numeric vector

Ba_INAA a numeric vector

Be a numeric vector

Bi a numeric vector

Br_IC a numeric vector

Br_INAA a numeric vector

Ca a numeric vector

Ca_INAA a numeric vector

CaO a numeric vector

Cd a numeric vector

Ce_INAA a numeric vector

Cl IC a numeric vector

Co a numeric vector

Co_INAA a numeric vector

EC a numeric vector

Cr a numeric vector

Cr_INAA a numeric vector

Cs INAA a numeric vector

Cu a numeric vector

Eu_INAA a numeric vector

F_IC a numeric vector

Fe a numeric vector

Fe_INAA a numeric vector

Fe2O3 a numeric vector

Hf INAA a numeric vector

Hg a numeric vector

Hg_INAA a numeric vector

Ir_INAA a numeric vector

K a numeric vector

K2O a numeric vector

La a numeric vector

La_INAA a numeric vector

Li a numeric vector

LOI a numeric vector

Lu_INAA a numeric vector

wt_INAA a numeric vector

Mg a numeric vector

MgO a numeric vector

Mn a numeric vector

MnO a numeric vector

Mo a numeric vector

Mo_INAA a numeric vector

Na a numeric vector

Na_INAA a numeric vector

Na2O a numeric vector

Nd_INAA a numeric vector

Ni a numeric vector

Ni_INAA a numeric vector

NO3_IC a numeric vector

P a numeric vector

P2O5 a numeric vector

Pb a numeric vector

pH a numeric vector

PO4_IC a numeric vector

Rb a numeric vector

S a numeric vector

Sb a numeric vector

Sb_INAA a numeric vector

Sc a numeric vector

Sc_INAA a numeric vector

Se a numeric vector

Se_INAA a numeric vector

Si a numeric vector

SiO2 a numeric vector

Sm_INAA a numeric vector

Sn_INAA a numeric vector

SO4_IC a numeric vector

Sr a numeric vector

Sr_INAA a numeric vector

SUM_XRF a numeric vector

Ta_INAA a numeric vector

Tb_INAA a numeric vector

Te a numeric vector

Th a numeric vector

Th_INAA a numeric vector

Ti a numeric vector

TiO2 a numeric vector

U_INAA a numeric vector

V a numeric vector

W_INAA a numeric vector

Y a numeric vector

Yb_INAA a numeric vector

Zn a numeric vector

Zn_INAA a numeric vector

ELEV a numeric vector

*COUN a numeric vector

*ASP a numeric vector

TOPC a numeric vector

LITO a numeric vector

Al_XRF a numeric vector

Ca_XRF a numeric vector

Fe_XRF a numeric vector

K_XRF a numeric vector

Mg_XRF a numeric vector

Mn_XRF a numeric vector

Na_XRF a numeric vector

P_XRF a numeric vector

Si_XRF a numeric vector

Ti_XRF a numeric vector

Note

For a more detailed description of this data set, see the help file chorizon in package mvoutlier.

Source

Kola Project (1993-1998)

References

Reimann, C., Filzmoser, P., Garrett, R.G. and Dutter, R. (2008) *Statistical Data Analysis Explained: Applied Environmental Statistics with R.* Wiley.

Examples

```
data(chorizonDL, package = "VIM")
summary(chorizonDL)
```

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colic

Colic horse data set

Description

This is a modified version of the original training data set taken from the UCI repository, see reference. The modifications are only related to having appropriate levels for factor variables. This data set is about horse diseases where the task is to determine, if the lesion of the horse was surgical or not.

Format

A training data frame with 300 observations on the following 31 variables.

surgery yes or no

age 1 equals an adult horse, 2 is a horse younger than 6 months

hospitalID ID

temp_rectal rectal temperature

pulse heart rate in beats per minute

respiratory_rate a normal rate is between 8 and 10

temp_extreme temperature of extremities

pulse_peripheral factor with four categories

capillayr_refill_time a clinical judgement. The longer the refill, the poorer the circulation. Possible values are 1 = < 3 seconds and 2 = >= 3 seconds

pain a subjective judgement of the horse's pain level

peristalsis an indication of the activity in the horse's gut. As the gut becomes more distended or the horse becomes more toxic, the activity decreases

abdominal_distension An animal with abdominal distension is likely to be painful and have reduced gut motility. A horse with severe abdominal distension is likely to require surgery just tio relieve the pressure

nasogastric_tube This refers to any gas coming out of the tube. A large gas cap in the stomach is likely to give the horse discomfort

nasogastric_reflux possible values are 1 = none, 2 = > 1 liter, 3 = < 1 liter. The greater amount of reflux, the more likelihood that there is some serious obstruction to the fluid passage from the rest of the intestine

nasogastric_reflux_PH scale is from 0 to 14 with 7 being neutral. Normal values are in the 3 to 4 range

rectal_examination Rectal examination. Absent feces probably indicates an obstruction

abdomen abdomen. possible values 1 = normal, 2 = other, 3 = firm feces in the large intestine, 4 = distended small intestine, 5 = distended large intestine

cell_volume packed cell volume. normal range is 30 to 50. The level rises as the circulation becomes compromised or as the animal becomes dehydrated.

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protein total protein. Normal values lie in the 6-7.5 (gms/dL) range. The higher the value the greater the dehydration

abdominocentesis_appearance Abdominocentesis appearance. A needle is put in the horse's abdomen and fluid is obtained from the abdominal cavity

abdomcentesis_protein abdomcentesis total protein. The higher the level of protein the more likely it is to have a compromised gut. Values are in gms/dL

outcome What eventually happened to the horse?

surgical_lesion retrospectively, was the problem (lesion) surgical?

lesion_type1 type of lesion

lesion_type2 type of lesion

lesion_type3 type of lesion

cp_data

temp_extreme_ordered temperature of extremities (ordered)

mucous_membranes_col mucous membranes. A subjective measurement of colour

mucous_membranes_group different recodings of mucous membrances

Source

https://archive.ics.uci.edu/ml/datasets/Horse+Colic Creators: Mary McLeish & Matt Cecile, Department of Computer Science, University of Guelph, Guelph, Ontario, Canada N1G 2W1 Donor: Will Taylor

Examples

data(colic)
aggr(colic)

collisions

Subset of the collision data

Description

Subset of the collision data from December 20. to December 31. 2018 from NYCD.

Details

Each record represents a collision in NYC by city, borough, precinct and cross street.

Source

https://data.cityofnewyork.us/Public-Safety/NYPD-Motor-Vehicle-Collisions/h9gi-nx95

Examples

```
data(collisions)
aggr(collisions)
```

colormapMiss 21

 ${\tt colormapMiss}$

Colored map with information about missing/imputed values

Description

Colored map in which the proportion or amount of missing/imputed values in each region is coded according to a continuous or discrete color scheme. The sequential color palette may thereby be computed in the *HCL* or the *RGB* color space.

Usage

```
colormapMiss(
  Х,
  region,
 map,
  imp_index = NULL,
 prop = TRUE,
 polysRegion = 1:length(x),
  range = NULL,
  n = NULL,
  col = c("red", "orange"),
  gamma = 2.2,
  fixup = TRUE,
  coords = NULL,
  numbers = TRUE,
  digits = 2,
  cex.numbers = 0.8,
  col.numbers = par("fg"),
  legend = TRUE,
  interactive = TRUE,
)
colormapMissLegend(
  xleft,
 ybottom,
  xright,
 ytop,
  cmap,
  n = 1000,
  horizontal = TRUE,
  digits = 2,
  cex.numbers = 0.8,
  col.numbers = par("fg"),
)
```

22 colormapMiss

Arguments

a numeric vector.

region a vector or factor of the same length as x giving the regions.

an object of any class that contains polygons and provides its own plot method map

(e.g., "SpatialPolygons" from package sp).

a logical-vector indicating which values of 'x' have been imputed. If given, it is imp_index

used for highlighting and the colors are adjusted according to the given colors

for imputed variables (see col).

prop a logical indicating whether the proportion of missing/imputed values should be

used rather than the total amount.

a numeric vector specifying the region that each polygon belongs to. polysRegion

a numeric vector of length two specifying the range (minimum and maximum) range

of the proportion or amount of missing/imputed values to be used for the color

scheme.

for colormapMiss, the number of equally spaced cut-off points for a discretized n

> color scheme. If this is not a positive integer, a continuous color scheme is used (the default). In the latter case, the number of rectangles to be drawn in the legend can be specified in colormapMissLegend. A reasonably large number

makes it appear continuously.

col the color range (start end end) to be used. RGB colors may be specified as

character strings or as objects of class "colorspace::RGB()". HCL colors need to be specified as objects of class "colorspace::polarLUV()". If only one color is supplied, it is used as end color, while the start color is taken to be

transparent for RGB or white for HCL.

numeric; the display *gamma* value (see colorspace::hex()). gamma

a logical indicating whether the colors should be corrected to valid RGB values fixup

(see colorspace::hex()).

coords a matrix or data. frame with two columns giving the coordinates for the labels. numbers

a logical indicating whether the corresponding proportions or numbers of miss-

ing/imputed values should be used as labels for the regions.

digits the number of digits to be used in the labels (in case of proportions).

cex.numbers the character expansion factor to be used for the labels.

col.numbers the color to be used for the labels.

legend a logical indicating whether a legend should be plotted.

a logical indicating whether more detailed information about missing/imputed interactive

values should be displayed interactively (see 'Details').

further arguments to be passed to plot.

xleft left x position of the legend. vbottom bottom y position of the legend. xright right x position of the legend. top y position of the legend. ytop

colormapMiss 23

cmap a list as returned by colormapMiss that contains the required information for

the legend.

horizontal a logical indicating whether the legend should be drawn horizontally or verti-

cally.

Details

The proportion or amount of missing/imputed values in x of each region is coded according to a continuous or discrete color scheme in the color range defined by col. In addition, the proportions or numbers can be shown as labels in the regions.

If interactive is TRUE, clicking in a region displays more detailed information about missing/imputed values on the console. Clicking outside the borders quits the interactive session.

Value

colormapMiss returns a list with the following components:

- nmiss a numeric vector containing the number of missing/imputed values in each region.
- nobs a numeric vector containing the number of observations in each region.
- pmiss a numeric vector containing the proportion of missing values in each region.
- prop a logical indicating whether the proportion of missing/imputed values have been used rather than the total amount.
- range the range of the proportion or amount of missing/imputed values corresponding to the color range.
- n either a positive integer giving the number of equally spaced cut-off points for a discretized color scheme, or NULL for a continuous color scheme.
- start the start color of the color scheme.
- end the end color of the color scheme.
- space a character string giving the color space (either "rgb" for RGB colors or "hcl" for HCL colors).
- gamma numeric; the display gamma value (see colorspace::hex()).
- fixup a logical indicating whether the colors have been corrected to valid RGB values (see colorspace::hex()).

Note

Some of the argument names and positions have changed with versions 1.3 and 1.4 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments cex.text and col.text can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use cex.numbers and col.numbers instead.

Author(s)

Andreas Alfons, modifications to show imputed values by Bernd Prantner

24 colSequence

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
colSequence(), growdotMiss(), mapMiss()
```

colSequence

HCL and RGB color sequences

Description

Compute color sequences by linear interpolation based on a continuous color scheme between certain start and end colors. Color sequences may thereby be computed in the *HCL* or *RGB* color space.

Usage

```
colSequence(p, start, end, space = c("hcl", "rgb"), ...)
colSequenceRGB(p, start, end, fixup = TRUE, ...)
colSequenceHCL(p, start, end, fixup = TRUE, ...)
```

Arguments

p	a numeric vector with values between 0 and 1 giving values to be used for interpolation between the start and end color (0 corresponds to the start color, 1 to the end color).
start, end	the start and end color, respectively. For HCL colors, each can be supplied as a vector of length three (hue, chroma, luminance) or an object of class "colorspace::polarLUV()". For RGB colors, each can be supplied as a character string, a vector of length three (red, green, blue) or an object of class "colorspace::RGB()".
space	character string; if start and end are both numeric, this determines whether they refer to HCL or RGB values. Possible values are "hcl" (for the HCL space) or "rgb" (for the RGB space).
	for colSequence, additional arguments to be passed to colSequenceHCL or colSequenceRGB. For colSequenceHCL and colSequenceRGB, additional arguments to be passed to colorspace::hex().
fixup	a logical indicating whether the colors should be corrected to valid RGB values (see colorspace::hex()).

Value

A character vector containing hexadecimal strings of the form "#RRGGBB".

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Author(s)

Andreas Alfons

References

Zeileis, A., Hornik, K., Murrell, P. (2009) Escaping RGBland: Selecting colors for statistical graphics. *Computational Statistics & Data Analysis*, **53** (9), 1259–1270.

See Also

```
colorspace::hex(), colorspace::sequential_hcl()
```

Examples

```
p <- c(0, 0.3, 0.55, 0.8, 1)
## HCL colors
colSequence(p, c(0, 0, 100), c(0, 100, 50))
colSequence(p, polarLUV(L=90, C=30, H=90), c(0, 100, 50))
## RGB colors
colSequence(p, c(1, 1, 1), c(1, 0, 0), space="rgb")
colSequence(p, RGB(1, 1, 0), "red")</pre>
```

countInf

Count number of infinite or missing values

Description

Count the number of infinite or missing values in a vector.

a vector.

Usage

```
countInf(x)
```

Arguments

X

Value

countInf returns the number of infinite values in x. countNA returns the number of missing values in x.

Author(s)

Andreas Alfons

26 diabetes

Examples

data(sleep, package="VIM")
countInf(log(sleep\$Dream))
countNA(sleep\$Dream)

diabetes

Indian Prime Diabetes Data

Description

The datasets consists of several medical predictor variables and one target variable, Outcome. Predictor variables includes the number of pregnancies the patient has had, their BMI, insulin level, age, and so on.

Format

A data frame with 768 observations on the following 9 variables.

Pregnancies Number of times pregnant

Glucose Plasma glucose concentration a 2 hours in an oral glucose tolerance test

BloodPressure Diastolic blood pressure (mm Hg) **SkinThickness** Triceps skin fold thickness (mm)

Insulin 2-Hour serum insulin (mu U/ml)

BMI Body mass index (weight in kg/(height in m)^2)

DiabetesPedigreeFunction Diabetes pedigree function

Age Age in years

Outcome Diabetes (yes or no)

Details

This dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether or not a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old of Pima Indian heritage.

Source

https://www.kaggle.com/uciml/pima-indians-diabetes-database/data

References

Smith, J.W., Everhart, J.E., Dickson, W.C., Knowler, W.C., & Johannes, R.S. (1988). Using the ADAP learning algorithm to forecast the onset of diabetes mellitus. In Proceedings of the Symposium on Computer Applications and Medical Care (pp. 261–265). IEEE Computer Society Press.

evaluation 27

Examples

```
data(diabetes)
aggr(diabetes)
```

evaluation

Error performance measures

Description

Various error measures evaluating the quality of imputations

Usage

```
evaluation(x, y, m, vartypes = "guess")
nrmse(x, y, m)
pfc(x, y, m)
msecov(x, y)
msecor(x, y)
```

Arguments

x matrix or data frame

y matrix or data frame of the same size as x

m the indicator matrix for missing cells

vartypes a vector of length ncol(x) specifying the variables types, like factor or numeric

Details

This function has been mainly written for procudures that evaluate imputation or replacement of rounded zeros. The ni parameter can thus, e.g. be used for expressing the number of rounded zeros.

Value

the error measures value

Author(s)

Matthias Templ

28 food

References

M. Templ, A. Kowarik, P. Filzmoser (2011) Iterative stepwise regression imputation using standard and robust methods. *Journal of Computational Statistics and Data Analysis*, Vol. 55, pp. 2793-2806.

Examples

```
data(iris)
iris_orig <- iris_imp <- iris
iris_imp$Sepal.Length[sample(1:nrow(iris), 10)] <- NA
iris_imp$Sepal.Width[sample(1:nrow(iris), 10)] <- NA
iris_imp$Species[sample(1:nrow(iris), 10)] <- NA
m <- is.na(iris_imp)
iris_imp <- kNN(iris_imp, imp_var = FALSE)
evaluation(iris_orig, iris_imp, m = m, vartypes = c(rep("numeric", 4), "factor"))
msecov(iris_orig[, 1:4], iris_imp[, 1:4])</pre>
```

food

Food consumption

Description

The relative consumption of certain food items in European and Scandinavian countries.

Format

A data frame with 16 observations on the following 21 variables.

Details

The numbers represent the percentage of the population consuming that food type.

Source

```
https://openmv.net/info/food-consumption
```

Examples

```
data(food)
str(food)
aggr(food)
```

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gapMiss

Missing value gap statistics

Description

Computes the average missing value gap of a vector.

Usage

```
gapMiss(x, what = mean)
```

Arguments

x a numeric vector

what default is the arithmetic mean. One can include an own function that returns a

vector of lenght 1 (e.g. median)

Details

The length of each sequence of missing values (gap) in a vector is calculated and the mean gap is reported

Value

The gap statistics

Author(s)

Matthias Templ based on a suggestion and draft from Huang Tian Yuan.

Examples

```
v <- rnorm(20)
v[3] <- NA
v[6:9] <- NA
v[13:17] <- NA
v
gapMiss(v)
gapMiss(v, what = median)
gapMiss(v, what = function(x) mean(x, trim = 0.1))
gapMiss(v, what = var)</pre>
```

30 gowerD

gowerD

Computes the extended Gower distance of two data sets

Description

The function gowerD is used by kNN to compute the distances for numerical, factor ordered and semi-continous variables.

Usage

```
gowerD(
  data.x,
  data.y = data.x,
  weights = rep(1, ncol(data.x)),
  numerical = colnames(data.x),
  factors = vector(),
  orders = vector(),
  mixed = vector(),
  levOrders = vector(),
  mixed.constant = rep(0, length(mixed)),
  returnIndex = FALSE,
  nMin = 1L,
  returnMin = FALSE,
  methodStand = "range"
)
```

Arguments

data.x	data frame
data.y	data frame
weights	numeric vector providing weights for the observations in x
numerical	names of numerical variables
factors	names of factor variables
orders	names of ordered variables
mixed	names of mixed variables
levOrders	vector with number of levels for each orders variable
mixed.constant	vector with length equal to the number of semi-continuous variables specifying the point of the semi-continuous distribution with non-zero probability
returnIndex	logical if TRUE return the index of the minimum distance
nMin	integer number of values with smallest distance to be returned
returnMin	logical if the computed distances for the indices should be returned
methodStand	character either "range" or "iqr", iqr is more robust for outliers

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Details

returnIndex=FALSE: a numerical matrix n x m with the computed distances returnIndex=TRUE: a named list with "ind" containing the requested indices and "mins" the computed distances

Examples

```
data(sleep)
# all variables used as numerical
gowerD(sleep)

# split in numerical an
gowerD(sleep, numerical = c("BodyWgt", "BrainWgt", "NonD", "Dream", "Sleep", "Span", "Gest"),
    orders = c("Pred", "Exp", "Danger"), levOrders = c(5,5,5))

# as before but only returning the index of the closest observation
gowerD(sleep, numerical = c("BodyWgt", "BrainWgt", "NonD", "Dream", "Sleep", "Span", "Gest"),
    orders = c("Pred", "Exp", "Danger"), levOrders = c(5,5,5), returnIndex = TRUE)
```

growdotMiss

Growing dot map with information about missing/imputed values

Description

Map with dots whose sizes correspond to the values in a certain variable. Observations with missing/imputed values in additional variables are highlighted.

Usage

```
growdotMiss(
  Х,
  coords,
 map,
  pos = 1,
  delimiter = NULL,
  selection = c("any", "all"),
  log = FALSE,
  col = c("skyblue", "red", "skyblue4", "red4", "orange", "orange4"),
  border = par("bg"),
  alpha = NULL,
  scale = NULL,
  size = NULL,
  exp = c(0, 0.95, 0.05),
  col.map = grey(0.5),
  legend = TRUE,
  legtitle = "Legend",
  cex.legtitle = par("cex"),
  cex.legtext = par("cex"),
  ncircles = 6,
```

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```
ndigits = 1,
interactive = TRUE,
...
)
```

Arguments

x a vector, matrix or data. frame.

coords a matrix or data. frame with two columns giving the spatial coordinates of the

observations.

map a background map to be passed to bgmap().

pos a numeric value giving the index of the variable determining the dot sizes.

delimiter a character-vector to distinguish between variables and imputation-indices for

imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

selection the selection method for highlighting missing/imputed values in multiple addi-

tional variables. Possible values are "any" (highlighting of missing/imputed values in *any* of the additional variables) and "all" (highlighting of missing/imputed

values in *all* of the additional variables).

log a logical indicating whether the variable given by pos should be log-transformed.

a vector of length six giving the colors to be used in the plot. If only one color is

supplied, it is used for the borders of non-highlighted dots and the surface area

of highlighted dots. Else if two colors are supplied, they are recycled.

border a vector of length four giving the colors to be used for the borders of the growing

dots. Use NA to omit borders.

alpha a numeric value between 0 and 1 giving the level of transparency of the colors,

or NULL. This can be used to prevent overplotting.

scale scaling factor of the map.

size a vector of length two giving the sizes for the smallest and largest dots.

exp a vector of length three giving the factors that define the shape of the exponential

function (see 'Details').

col.map the color to be used for the background map.

legend a logical indicating whether a legend should be plotted.

legtitle the title for the legend.

cex.legtitle the character expansion factor to be used for the title of the legend.

cex.legtext the character expansion factor to be used in the legend.

ncircles the number of circles displayed in the legend.

ndigits the number of digits displayed in the legend. Note that \ this is just a suggestion

(see format()).

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interactive	a logical indicating whether information about certain observations can be displayed interactively (see 'Details').
•••	for growdotMiss, further arguments and graphical parameters to be passed to bgmap(). For bubbleMiss, the arguments to be passed to growdotMiss.

Details

The smallest dots correspond to the 10\ the 99\ defining the shape of the exponential function. Missings/imputed missings in the variable of interest will be drawn as rectangles.

If interactive=TRUE, detailed information for an observation can be printed on the console by clicking on the corresponding point. Clicking in a region that does not contain any points quits the interactive session.

Note

The function was renamed to growdotMiss in version 1.3. bubbleMiss is a (deprecated) wrapper for growdotMiss for back compatibility with older versions. However, due to extended functionality, some of the argument positions have changed.

The code is based on (removed from CRAN) bubbleFIN from package StatDA.

Author(s)

Andreas Alfons, Matthias Templ, Peter Filzmoser, Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
bgmap(), mapMiss(), colormapMiss()
```

Examples

```
data(chorizonDL, package = "VIM")
data(kola.background, package = "VIM")
coo <- chorizonDL[, c("XCOO", "YCOO")]
## for missing values
x <- chorizonDL[, c("Ca","As", "Bi")]
growdotMiss(x, coo, kola.background, border = "white")

## for imputed values
x_imp <- kNN(chorizonDL[,c("Ca","As","Bi")])
growdotMiss(x_imp, coo, kola.background, delimiter = "_imp", border = "white")</pre>
```

34 histMiss

histMiss

Histogram with information about missing/imputed values

Description

Histogram with highlighting of missing/imputed values in other variables by splitting each bin into two parts. Additionally, information about missing/imputed values in the variable of interest is shown on the right hand side.

Usage

```
histMiss(
  delimiter = NULL,
  pos = 1,
  selection = c("any", "all"),
  breaks = "Sturges",
  right = TRUE,
  col = c("skyblue", "red", "skyblue4", "red4", "orange", "orange4"),
  border = NULL,
  main = NULL,
  sub = NULL,
  xlab = NULL,
  ylab = NULL,
  axes = TRUE,
  only.miss = TRUE,
  miss.labels = axes,
  interactive = TRUE,
)
```

Arguments

a vector, matrix or data. frame.

delimiter a character-vector to distinguish between variables and imputation-indices for

> imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

a numeric value giving the index of the variable of interest. Additional variables pos

in x are used for highlighting.

selection the selection method for highlighting missing/imputed values in multiple additional variables. Possible values are "any" (highlighting of missing/imputed val-

ues in any of the additional variables) and "all" (highlighting of missing/imputed

values in all of the additional variables).

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breaks	either a character string naming an algorithm to compute the breakpoints (see hist()), or a numeric value giving the number of cells.
right	logical; if TRUE, the histogram cells are right-closed (left-open) intervals.
col	a vector of length six giving the colors to be used. If only one color is supplied, the bars are transparent and the supplied color is used for highlighting missing/imputed values. Else if two colors are supplied, they are recycled.
border	the color to be used for the border of the cells. Use border=NA to omit borders.
main, sub	main and sub title.
xlab, ylab	axis labels.
axes	a logical indicating whether axes should be drawn on the plot.
only.miss	logical; if TRUE, the missing/imputed values in the first variable are visualized by a single bar. Otherwise, a small barplot is drawn on the right hand side (see 'Details').
miss.labels	either a logical indicating whether label(s) should be plotted below the bar(s) on the right hand side, or a character string or vector giving the label(s) (see 'Details').
interactive	a logical indicating whether the variables can be switched interactively (see 'Details').
	further graphical parameters to be passed to graphics::title() and graphics::axis().

Details

If more than one variable is supplied, the bins for the variable of interest will be split according to missingness/number of imputed missings in the additional variables.

If only.miss=TRUE, the missing/imputed values in the variable of interest are visualized by one bar on the right hand side. If additional variables are supplied, this bar is again split into two parts according to missingness/number of imputed missings in the additional variables.

Otherwise, a small barplot consisting of two bars is drawn on the right hand side. The first bar corresponds to observed values in the variable of interest and the second bar to missing/imputed values. Since these two bars are not on the same scale as the main barplot, a second y-axis is plotted on the right (if axes=TRUE). Each of the two bars are again split into two parts according to missingness/number of imputed missings in the additional variables. Note that this display does not make sense if only one variable is supplied, therefore only.miss is ignored in that case.

If interactive=TRUE, clicking in the left margin of the plot results in switching to the previous variable and clicking in the right margin results in switching to the next variable. Clicking anywhere else on the graphics device quits the interactive session. When switching to a categorical variable, a barplot is produced rather than a histogram.

Value

a list with the following components:

- breaks the breakpoints.
- counts the number of observations in each cell.
- missings the number of highlighted observations in each cell.
- mids the cell midpoints.

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Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments axisnames and names.miss can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use miss.labels instead.

Author(s)

Andreas Alfons, Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
spineMiss(), barMiss()
Other plotting functions: aggr(), barMiss(), marginmatrix(), marginplot(), matrixplot(),
mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

```
data(tao, package = "VIM")
## for missing values
x <- tao[, c("Air.Temp", "Humidity")]
histMiss(x)
histMiss(x, only.miss = FALSE)

## for imputed values
x_IMPUTED <- kNN(tao[, c("Air.Temp", "Humidity")])
histMiss(x_IMPUTED, delimiter = "_imp")
histMiss(x_IMPUTED, delimiter = "_imp", only.miss = FALSE)</pre>
```

hotdeck

Hot-Deck Imputation

Description

Implementation of the popular Sequential, Random (within a domain) hot-deck algorithm for imputation.

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Usage

```
hotdeck(
  data,
  variable = NULL,
  ord_var = NULL,
  domain_var = NULL,
  makeNA = NULL,
  NAcond = NULL,
  impNA = TRUE,
  donorcond = NULL,
  imp_var = TRUE,
  imp_suffix = "imp"
)
```

Arguments

data	data.frame or matrix
variable	variables where missing values should be imputed (not overlapping with ord_var)
ord_var	variables for sorting the data set before imputation (not overlapping with variable)
domain_var	variables for building domains and impute within these domains
makeNA	list of length equal to the number of variables, with values, that should be converted to NA for each variable
NAcond	list of length equal to the number of variables, with a condition for imputing a NA
impNA	TRUE/FALSE whether NA should be imputed
donorcond	list of length equal to the number of variables, with a donorcond condition as character string. e.g. ">5" or c(">5","<10). If the list element for a variable is NULL no condition will be applied for this variable.
imp_var	TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be created show the imputation status
imp_suffix	suffix for the TRUE/FALSE variables showing the imputation status

Value

the imputed data set.

Note

If the sequential hotdeck does not lead to a suitable, a random donor in the group will be used.

Author(s)

Alexander Kowarik

38 impPCA

References

A. Kowarik, M. Templ (2016) Imputation with R package VIM. *Journal of Statistical Software*, 74(7), 1-16.

See Also

```
Other imputation methods: impPCA(), irmi(), kNN(), matchImpute(), medianSamp(), rangerImpute(), regressionImp(), sampleCat()
```

Examples

```
data(sleep)
sleepI <- hotdeck(sleep)</pre>
sleepI2 <- hotdeck(sleep,ord_var="BodyWgt",domain_var="Pred")</pre>
# Usage of donorcond in a simple example
sleepI3 <- hotdeck(</pre>
  sleep,
  variable = c("NonD", "Dream", "Sleep", "Span", "Gest"),
  ord_var = "BodyWgt", domain_var = "Pred",
  donorcond = list(">4", "<17", ">1.5", "%between%c(8,13)", ">5")
set.seed(132)
nRows <- 1e3
# Generate a data set with nRows rows and several variables
x <- data.frame(</pre>
  x = rnorm(nRows), y = rnorm(nRows),
  z = sample(LETTERS, nRows, replace = TRUE),
  d1 = sample(LETTERS[1:3], nRows, replace = TRUE),
  d2 = sample(LETTERS[1:2], nRows, replace = TRUE),
  o1 = rnorm(nRows), o2 = rnorm(nRows), o3 = rnorm(100)
)
origX <- x
x[sample(1:nRows,nRows/10), 1] <- NA
x[sample(1:nRows,nRows/10), 2] <- NA
x[sample(1:nRows,nRows/10), 3] <- NA
x[sample(1:nRows,nRows/10), 4] <- NA
xImp \leftarrow hotdeck(x,ord\_var = c("o1", "o2", "o3"), domain\_var = "d2")
```

impPCA

Iterative EM PCA imputation

Description

Greedy algorithm for EM-PCA including robust methods

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Usage

```
impPCA(
    x,
    method = "classical",
    m = 1,
    eps = 0.5,
    k = ncol(x) - 1,
    maxit = 100,
    boot = FALSE,
    verbose = TRUE
)
```

Arguments

X	data.frame or matrix
method	"classical" or "mcd" (robust estimation)
m	number of multiple imputations (only if parameter boot equals TRUE)
eps	threshold for convergence
k	number of principal components for reconstruction of x
maxit	maximum number of iterations
boot	residual bootstrap (if TRUE)
verbose	TRUE/FALSE if additional information about the imputation process should be printed

Value

the imputed data set. If boot = FALSE this is a data.frame. If boot = TRUE this is a list where each list element contains a data.frame.

Author(s)

Matthias Templ

References

Serneels, Sven and Verdonck, Tim (2008). Principal component analysis for data containing outliers and missing elements. Computational Statistics and Data Analysis, Elsevier, vol. 52(3), pages 1712-1727

See Also

```
Other imputation methods: hotdeck(), irmi(), kNN(), matchImpute(), medianSamp(), rangerImpute(), regressionImp(), sampleCat()
```

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Examples

```
data(Animals, package = "MASS")
Animals$brain[19] <- Animals$brain[19] + 0.01</pre>
Animals <- log(Animals)</pre>
colnames(Animals) <- c("log(body)", "log(brain)")</pre>
Animals_na <- Animals
probs <- abs(Animals$`log(body)`^2)</pre>
probs <- rep(0.5, nrow(Animals))</pre>
probs[c(6,16,26)] <- 0
set.seed(1234)
Animals_na[sample(1:nrow(Animals), 10, prob = probs), "log(brain)"] <- NA</pre>
w <- is.na(Animals_na$`log(brain)`)</pre>
impPCA(Animals_na)
impPCA(Animals_na, method = "mcd")
impPCA(Animals_na, boot = TRUE, m = 10)
impPCA(Animals_na, method = "mcd", boot = TRUE)[[1]]
plot(`log(brain)` ~ `log(body)`, data = Animals, type = "n", ylab = "", xlab="")
mtext(text = "impPCA robust", side = 3)
points(Animals$`log(body)`[!w], Animals$`log(brain)`[!w])
points(Animals$`log(body)`[w], Animals$`log(brain)`[w], col = "grey", pch = 17)
imputed <- impPCA(Animals_na, method = "mcd", boot = TRUE)[[1]]</pre>
colnames(imputed) <- c("log(body)", "log(brain)")</pre>
points(imputed$`log(body)`[w], imputed$`log(brain)`[w], col = "red", pch = 20, cex = 1.4)
segments(x0 = Animals$`log(body)`[w], x1 = imputed$`log(body)`[w], y0 = Animals$`log(brain)`[w],
y1 = imputed$`log(brain)`[w], lty = 2, col = "grey")
legend("topleft", legend = c("non-missings", "set to missing", "imputed values"),
pch = c(1,17,20), col = c("black","grey","red"), cex = 0.7)
mape <- round(100* 1/sum(is.na(Animals_na$`log(brain)`)) * sum(abs((Animals$`log(brain)` -</pre>
imputed$`log(brain)`) / Animals$`log(brain)`)), 2)
s2 <- var(Animals$`log(brain)`)</pre>
nrmse <- round(sqrt(1/sum(is.na(Animals_na$`log(brain)`)) * sum(abs((Animals$`log(brain)` -</pre>
imputed$`log(brain)`) / s2))), 2)
text(x = 8, y = 1.5, labels = paste("MAPE =", mape))
text(x = 8, y = 0.5, labels = paste("NRMSE =", nrmse))
```

initialise

Initialization of missing values

Description

Rough estimation of missing values in a vector according to its type.

```
initialise(x, mixed, method = "kNN", mixed.constant = NULL)
```

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Arguments

X	a vector.
mixed	a character vector containing the names of variables of type mixed (semi-continous).
method	Method used for Initialization (median or kNN)
mixed.constant	vector with length equal to the number of semi-continuous variables specifying

the point of the semi-continuous distribution with non-zero probability

Details

Missing values are imputed with the mean for vectors of class "numeric", with the median for vectors of class "integer", and with the mode for vectors of class "factor". Hence, x should be prepared in the following way: assign class "numeric" to numeric vectors, assign class "integer" to ordinal vectors, and assign class "factor" to nominal or binary vectors.

Value

the initialized vector.

Note

The function is used internally by some imputation algorithms.

Author(s)

Matthias Templ, modifications by Andreas Alfons

irmi

Iterative robust model-based imputation (IRMI)

Description

In each step of the iteration, one variable is used as a response variable and the remaining variables serve as the regressors.

```
irmi(
    x,
    eps = 5,
    maxit = 100,
    mixed = NULL,
    mixed.constant = NULL,
    count = NULL,
    step = FALSE,
    robust = FALSE,
    takeAll = TRUE,
```

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```
noise = TRUE,
noise.factor = 1,
force = FALSE,
robMethod = "MM",
force.mixed = TRUE,
mi = 1,
addMixedFactors = FALSE,
trace = FALSE,
init.method = "kNN",
modelFormulas = NULL,
multinom.method = "multinom",
imp_var = TRUE,
imp_suffix = "imp"
)
```

Arguments

x data.frame or matrix

eps threshold for convergency
maxit maximum number of iterations

mixed column index of the semi-continuous variables

mixed.constant vector with length equal to the number of semi-continuous variables specifying

the point of the semi-continuous distribution with non-zero probability

count column index of count variables

step a stepwise model selection is applied when the parameter is set to TRUE

robust if TRUE, robust regression methods will be applied

takeAll takes information of (initialised) missings in the response as well for regression

imputation.

noise irmi has the option to add a random error term to the imputed values, this creates

the possibility for multiple imputation. The error term has mean 0 and variance

corresponding to the variance of the regression residuals.

noise.factor amount of noise.

force if TRUE, the algorithm tries to find a solution in any case, possible by using

different robust methods automatically.

robMethod regression method when the response is continuous.

force.mixed if TRUE, the algorithm tries to find a solution in any case, possible by using

different robust methods automatically.

mi number of multiple imputations.

addMixedFactors

if TRUE add additional factor variable for each mixed variable as X variable in

the regression

trace Additional information about the iterations when trace equals TRUE.

init.method Method for initialization of missing values (kNN or median)

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modelFormulas a named list with the name of variables for the rhs of the formulas, which must

contain a rhs formula for each variable with missing values, it should look like 'list(y1=c("x1","x2"),y2=c("x1","x3"))" if factor variables for the mixed variables for the mixed variables.

ables should be created for the regression models

multinom.method

Method for estimating the multinomial models (current default and only avail-

able method is multinom)

imp_var TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be

created show the imputation status

imp_suffix suffix for the TRUE/FALSE variables showing the imputation status

Details

The method works sequentially and iterative. The method can deal with a mixture of continuous, semi-continuous, ordinal and nominal variables including outliers.

A full description of the method can be found in the mentioned reference.

Value

the imputed data set.

Author(s)

Matthias Templ, Alexander Kowarik

References

M. Templ, A. Kowarik, P. Filzmoser (2011) Iterative stepwise regression imputation using standard and robust methods. *Journal of Computational Statistics and Data Analysis*, Vol. 55, pp. 2793-2806.

A. Kowarik, M. Templ (2016) Imputation with R package VIM. *Journal of Statistical Software*, 74(7), 1-16.

See Also

```
mi::mi()
```

```
Other imputation methods: hotdeck(), impPCA(), kNN(), matchImpute(), medianSamp(), rangerImpute(), regressionImp(), sampleCat()
```

Examples

```
data(sleep)
irmi(sleep)

data(testdata)
imp_testdata1 <- irmi(testdata$wna, mixed = testdata$mixed)

# mixed.constant != 0 (-10)
testdata$wna$m1[testdata$wna$m1 == 0] <- -10</pre>
```

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```
testdata$wna$m2 <- log(testdata$wna$m2 + 0.001)
imp_testdata2 <- irmi(</pre>
 testdata$wna,
 mixed = testdata$mixed,
 mixed.constant = c(-10, log(0.001))
imp_testdata2$m2 <- exp(imp_testdata2$m2) - 0.001</pre>
#example with fixed formulas for the variables with missing
form = list(
 NonD = c("BodyWgt", "BrainWgt"),
 Dream = c("BodyWgt", "BrainWgt"),
 Sleep = c("BrainWgt"
 Span = c("BodyWgt")
 Gest = c("BodyWgt", "BrainWgt")
irmi(sleep, modelFormulas = form, trace = TRUE)
# Example with ordered variable
td <- testdata$wna
td$c1 <- as.ordered(td$c1)
irmi(td)
```

kNN

k-Nearest Neighbour Imputation

Description

k-Nearest Neighbour Imputation based on a variation of the Gower Distance for numerical, categorical, ordered and semi-continous variables.

```
kNN(
   data,
   variable = colnames(data),
   metric = NULL,
   k = 5,
   dist_var = colnames(data),
   weights = NULL,
   numFun = median,
   catFun = maxCat,
   makeNA = NULL,
   NAcond = NULL,
   impNA = TRUE,
   donorcond = NULL,
   mixed = vector(),
   mixed.constant = NULL,
```

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```
trace = FALSE,
imp_var = TRUE,
imp_suffix = "imp",
addRF = FALSE,
onlyRF = FALSE,
addRandom = FALSE,
useImputedDist = TRUE,
weightDist = FALSE,
methodStand = "range",
ordFun = medianSamp
```

Arguments

data data.frame or matrix

variable variables where missing values should be imputed metric metric to be used for calculating the distances between

k number of Nearest Neighbours used

dist_var names or variables to be used for distance calculation

weights weights for the variables for distance calculation. If weights = "auto" weights

will be selected based on variable importance from random forest regression, using function ranger::ranger(). Weights are calculated for each variable

seperately.

numFun function for aggregating the k Nearest Neighbours in the case of a numerical

variable

catFun function for aggregating the k Nearest Neighbours in the case of a categorical

variable

makeNA list of length equal to the number of variables, with values, that should be con-

verted to NA for each variable

NAcond list of length equal to the number of variables, with a condition for imputing a

NA

impNA TRUE/FALSE whether NA should be imputed

donorcond list of length equal to the number of variables, with a donorcond condition as

character string. e.g. a list element can be ">5" or c(">5","<10). If the list element for a variable is NULL no condition will be applied for this variable.

mixed names of mixed variables

mixed.constant vector with length equal to the number of semi-continuous variables specifying

the point of the semi-continuous distribution with non-zero probability

trace TRUE/FALSE if additional information about the imputation process should be

printed

imp_var TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be

created show the imputation status

imp_suffix suffix for the TRUE/FALSE variables showing the imputation status

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addRF TRUE/FALSE each variable will be modelled using random forest regression

(ranger::ranger()) and used as additional distance variable.

onlyRF TRUE/FALSE if TRUE only additional distance variables created from random

forest regression will be used as distance variables.

addRandom TRUE/FALSE if an additional random variable should be added for distance

calculation

useImputedDist TRUE/FALSE if an imputed value should be used for distance calculation for

imputing another variable. Be aware that this results in a dependency on the

ordering of the variables.

weightDist TRUE/FALSE if the distances of the k nearest neighbours should be used as

weights in the aggregation step

methodStand either "range" or "iqr" to be used in the standardization of numeric vaiables in

the gower distance

ordFun function for aggregating the k Nearest Neighbours in the case of a ordered factor

variable

Value

the imputed data set.

Author(s)

Alexander Kowarik, Statistik Austria

References

A. Kowarik, M. Templ (2016) Imputation with R package VIM. *Journal of Statistical Software*, 74(7), 1-16.

See Also

```
Other imputation methods: hotdeck(), impPCA(), irmi(), matchImpute(), medianSamp(), rangerImpute(), regressionImp(), sampleCat()
```

Examples

```
data(sleep)
kNN(sleep)
library(laeken)
kNN(sleep, numFun = weightedMean, weightDist=TRUE)
```

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kola.background

Background map for the Kola project data

Description

Coordinates of the Kola background map.

Source

```
Kola Project (1993-1998)
```

References

Reimann, C., Filzmoser, P., Garrett, R.G. and Dutter, R. (2008) *Statistical Data Analysis Explained: Applied Environmental Statistics with R.* Wiley, 2008.

Examples

```
data(kola.background, package = "VIM")
bgmap(kola.background)
```

mapMiss

Map with information about missing/imputed values

Description

Map of observed and missing/imputed values.

```
mapMiss(
    x,
    coords,
    map,
    delimiter = NULL,
    selection = c("any", "all"),
    col = c("skyblue", "red", "orange"),
    alpha = NULL,
    pch = c(19, 15),
    col.map = grey(0.5),
    legend = TRUE,
    interactive = TRUE,
    ...
)
```

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Arguments

a vector, matrix or data. frame. Х

a data. frame or matrix with two columns giving the spatial coordinates of the coords

observations.

a background map to be passed to bgmap(). map

delimiter a character-vector to distinguish between variables and imputation-indices for

> imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

the selection method for displaying missing/imputed values in the map. Possible selection

values are "any" (display missing/imputed values in any variable) and "all"

(display missing/imputed values in all variables).

col a vector of length three giving the colors to be used for observed, missing and

imputed values. If a single color is supplied, it is used for all values.

a numeric value between 0 and 1 giving the level of transparency of the colors, alpha

or NULL. This can be used to prevent overplotting.

pch a vector of length two giving the plot characters to be used for observed and

missing/imputed values. If a single plot character is supplied, it will be used for

both.

col.map the color to be used for the background map.

legend a logical indicating whether a legend should be plotted.

interactive a logical indicating whether information about selected observations can be dis-

played interactively (see 'Details').

further graphical parameters to be passed to bgmap() and graphics::points().

Details

If interactive=TRUE, detailed information for an observation can be printed on the console by clicking on the corresponding point. Clicking in a region that does not contain any points quits the interactive session.

Author(s)

Matthias Templ, Andreas Alfons, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. Journal of Advances in Data Analysis and Classification, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

bgmap(), bubbleMiss(), colormapMiss()

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Examples

```
data(chorizonDL, package = "VIM")
data(kola.background, package = "VIM")
coo <- chorizonDL[, c("XCOO", "YCOO")]
## for missing values
x <- chorizonDL[, c("As", "Bi")]
mapMiss(x, coo, kola.background)
## for imputed values
x_imp <- kNN(chorizonDL[, c("As", "Bi")])
mapMiss(x_imp, coo, kola.background, delimiter = "_imp")</pre>
```

marginmatrix

Marginplot Matrix

Description

Create a scatterplot matrix with information about missing/imputed values in the plot margins of each panel.

Usage

```
marginmatrix(
    x,
    delimiter = NULL,
    col = c("skyblue", "red", "red4", "orange", "orange4"),
    alpha = NULL,
    ...
)
```

Arguments

x

a matrix or data. frame.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

col

a vector of length five giving the colors to be used in the marginplots in the off-diagonal panels. The first color is used for the scatterplot and the boxplots for the available data, the second/fourth color for the univariate scatterplots and boxplots for the missing/imputed values in one variable, and the third/fifth color for the frequency of missing/imputed values in both variables (see 'Details'). If only one color is supplied, it is used for the bivariate and univariate scatterplots

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	and the boxplots for missing/imputed values in one variable, whereas the boxplots for the available data are transparent. Else if two colors are supplied, the second one is recycled.
alpha	a numeric value between 0 and 1 giving the level of transparency of the colors, or NULL. This can be used to prevent overplotting.
• • •	further arguments and graphical parameters to be passed to pairsVIM() and marginplot(). par("oma") will be set appropriately unless supplied (see graphics::par()).

Details

marginmatrix uses pairsVIM() with a panel function based on marginplot().

The graphical parameter oma will be set unless supplied as an argument.

Author(s)

Andreas Alfons, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
marginplot(), pairsVIM(), scattmatrixMiss()
Other plotting functions: aggr(), barMiss(), histMiss(), marginplot(), matrixplot(), mosaicMiss(),
pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(), spineMiss()
```

Examples

```
data(sleep, package = "VIM")
## for missing values
x <- sleep[, 1:5]
x[,c(1,2,4)] <- log10(x[,c(1,2,4)])
marginmatrix(x)

## for imputed values
x_imp <- kNN(sleep[, 1:5])
x_imp[,c(1,2,4)] <- log10(x_imp[,c(1,2,4)])
marginmatrix(x_imp, delimiter = "_imp")</pre>
```

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marginplot

Scatterplot with additional information in the margins

Description

In addition to a standard scatterplot, information about missing/imputed values is shown in the plot margins. Furthermore, imputed values are highlighted in the scatterplot.

Usage

```
marginplot(
  Χ,
 delimiter = NULL,
  col = c("skyblue", "red", "red4", "orange", "orange4"),
  alpha = NULL,
  pch = c(1, 16),
  cex = par("cex"),
  numbers = TRUE,
  cex.numbers = par("cex"),
  zeros = FALSE,
 xlim = NULL,
 ylim = NULL,
 main = NULL,
  sub = NULL,
  xlab = NULL,
 ylab = NULL,
  ann = par("ann"),
  axes = TRUE,
  frame.plot = axes,
)
```

Arguments

x

a matrix or data. frame with two columns.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

col

a vector of length five giving the colors to be used in the plot. The first color is used for the scatterplot and the boxplots for the available data. In case of missing values, the second color is taken for the univariate scatterplots and boxplots for missing values in one variable and the third for the frequency of missing/imputed values in both variables (see 'Details'). Otherwise, in case of imputed values,

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the fourth color is used for the highlighting, the frequency, the univariate scatterplot and the boxplots of mputed values in the first variable and the fifth color for the same applied to the second variable. A black color is used for the highlighting and the frequency of imputed values in both variables instead. If only one color is supplied, it is used for the bivariate and univariate scatterplots and the boxplots for missing/imputed values in one variable, whereas the boxplots for the available data are transparent. Else if two colors are supplied, the second one is recycled.

alpha a numeric value between 0 and 1 giving the level of transparency of the colors,

or NULL. This can be used to prevent overplotting.

pch a vector of length two giving the plot symbols to be used for the scatterplot and

the univariate scatterplots. If a single plot character is supplied, it is used for the scatterplot and the default value will be used for the univariate scatterplots (see

'Details').

cex the character expansion factor to be used for the bivariate and univariate scatter-

plots.

numbers a logical indicating whether the frequencies of missing/imputed values should

be displayed in the lower left of the plot (see 'Details').

cex.numbers the character expansion factor to be used for the frequencies of the missing/imputed

values.

zeros a logical vector of length two indicating whether the variables are semi-continuous,

i.e., contain a considerable amount of zeros. If TRUE, only the non-zero observations are used for drawing the respective boxplot. If a single logical is supplied,

it is recycled.

xlim, ylim axis limits.

main, sub main and sub title.

xlab, ylab axis labels.

ann a logical indicating whether plot annotation (main, sub, xlab, ylab) should be

displayed.

axes a logical indicating whether both axes should be drawn on the plot. Use graphi-

cal parameter "xaxt" or "yaxt" to suppress only one of the axes.

frame.plot a logical indicating whether a box should be drawn around the plot.

... further graphical parameters to be passed down (see graphics::par()).

Details

Boxplots for available and missing/imputed data, as well as univariate scatterplots for missing/imputed values in one variable are shown in the plot margins.

Imputed values in either of the variables are highlighted in the scatterplot.

Furthermore, the frequencies of the missing/imputed values can be displayed by a number (lower left of the plot). The number in the lower left corner is the number of observations that are missing/imputed in both variables.

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Note

Some of the argument names and positions have changed with versions 1.3 and 1.4 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the argument cex.text can still be supplied to ...{} and is handled correctly. Nevertheless, it is deprecated and no longer documented. Use cex.numbers instead.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
scattMiss()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), matrixplot(),
mosaicMiss(),pairsVIM(),parcoordMiss(),pbox(),scattJitt(),scattMiss(),scattmatrixMiss(),
```

Examples

spineMiss()

```
data(tao, package = "VIM")
data(chorizonDL, package = "VIM")
## for missing values
marginplot(tao[,c("Air.Temp", "Humidity")])
marginplot(log10(chorizonDL[,c("CaO", "Bi")]))

## for imputed values
marginplot(kNN(tao[,c("Air.Temp", "Humidity")]), delimiter = "_imp")
marginplot(kNN(log10(chorizonDL[,c("CaO", "Bi")])), delimiter = "_imp")
```

matchImpute

Fast matching/imputation based on categorical variable

Description

Suitable donors are searched based on matching of the categorical variables. The variables are dropped in reversed order, so that the last element of 'match_var' is dropped first and the first element of the vector is dropped last.

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Usage

```
matchImpute(
  data,
  variable = colnames(data)[!colnames(data) %in% match_var],
  match_var,
  imp_var = TRUE,
  imp_suffix = "imp"
)
```

Arguments

data data.frame, data.table or matrix

variable variables to be imputed match_var variables used for matching

imp_var TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be

created show the imputation status

imp_suffix suffix for the TRUE/FALSE variables showing the imputation status

Details

The method works by sampling values from the suitable donors.

Value

the imputed data set.

Author(s)

Johannes Gussenbauer, Alexander Kowarik

See Also

hotdeck()

```
Other imputation methods: hotdeck(), impPCA(), irmi(), kNN(), medianSamp(), rangerImpute(), regressionImp(), sampleCat()
```

Examples

```
data(sleep,package="VIM")
imp_data <- matchImpute(sleep,variable=c("NonD","Dream","Sleep","Span","Gest"),
    match_var=c("Exp","Danger"))

data(testdata,package="VIM")
imp_testdata1 <- matchImpute(testdata$wna,match_var=c("c1","c2","b1","b2"))

dt <- data.table::data.table(testdata$wna)
imp_testdata2 <- matchImpute(dt,match_var=c("c1","c2","b1","b2"))</pre>
```

matrixplot 55

matrixplot

Matrix plot

Description

Create a matrix plot, in which all cells of a data matrix are visualized by rectangles. Available data is coded according to a continuous color scheme, while missing/imputed data is visualized by a clearly distinguishable color.

Usage

```
matrixplot(
  delimiter = NULL,
  sortby = NULL,
  col = c("red", "orange"),
  fixup = TRUE,
  xlim = NULL,
 ylim = NULL,
 main = NULL,
  sub = NULL,
  xlab = NULL,
 ylab = NULL,
  axes = TRUE,
  labels = axes,
  xpd = NULL,
  interactive = TRUE,
)
```

Arguments

x a matrix or data. frame.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

sortby

a numeric or character value specifying the variable to sort the data matrix by, or NULL to plot without sorting.

col

the colors to be used in the plot. RGB colors may be specified as character strings or as objects of class "colorspace::RGB()". HCL colors need to be specified as objects of class "colorspace::polarLUV()". If only one color is supplied, it is used for missing and imputed data and a greyscale is used for available data. If two colors are supplied, the first is used for missing and the

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second for imputed data and a greyscale for available data. If three colors are supplied, the first is used as end color for the available data, while the start color is taken to be transparent for RGB or white for HCL. Missing/imputed data is visualized by the second/third color in this case. If four colors are supplied, the first is used as start color and the second as end color for the available data, while the third/fourth color is used for missing/imputed data.

fixup a logical indicating whether the colors should be corrected to valid RGB values

(see colorspace::hex()).

xlim, ylim axis limits.

main, sub main and sub title.

xlab, ylab axis labels.

axes a logical indicating whether axes should be drawn on the plot.

labels either a logical indicating whether labels should be plotted below each column,

or a character vector giving the labels.

xpd a logical indicating whether the rectangles should be allowed to go outside the

plot region. If NULL, it defaults to TRUE unless axis limits are specified.

interactive a logical indicating whether a variable to be used for sorting can be selected

interactively (see 'Details').

for matrixplot and iimagMiss, further graphical parameters to be passed to

graphics::plot.window(), graphics::title() and graphics::axis(). For

TKRmatrixplot, further arguments to be passed to matrixplot.

Details

In a *matrix plot*, all cells of a data matrix are visualized by rectangles. Available data is coded according to a continuous color scheme. To compute the colors via interpolation, the variables are first scaled to the interval between 0 and 1. Missing/imputed values can then be visualized by a clearly distinguishable color. It is thereby possible to use colors in the *HCL* or *RGB* color space. A simple way of visualizing the magnitude of the available data is to apply a greyscale, which has the advantage that missing/imputed values can easily be distinguished by using a color such as red/orange. Note that -Inf and Inf are always assigned the begin and end color, respectively, of the continuous color scheme.

Additionally, the observations can be sorted by the magnitude of a selected variable. If interactive is TRUE, clicking in a column redraws the plot with observations sorted by the corresponding variable. Clicking anywhere outside the plot region quits the interactive session.

Note

This is a much more powerful extension to the function imagmiss in the former CRAN package dprep.

iimagMiss is deprecated and may be omitted in future versions of VIM. Use matrixplot instead.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

maxCat 57

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(), spineMiss()
```

Examples

```
data(sleep, package = "VIM")
## for missing values
x <- sleep[, -(8:10)]
x[,c(1,2,4,6,7)] <- log10(x[,c(1,2,4,6,7)])
matrixplot(x, sortby = "BrainWgt")

## for imputed values
x_imp <- kNN(sleep[, -(8:10)])
x_imp[,c(1,2,4,6,7)] <- log10(x_imp[,c(1,2,4,6,7)])
matrixplot(x_imp, delimiter = "_imp", sortby = "BrainWgt")</pre>
```

maxCat

Aggregation function for a factor variable

Description

The function maxCat chooses the level with the most occurrences and random if the maximum is not unique.

Usage

```
maxCat(x, weights = NULL)
```

Arguments

x factor vector

weights numeric vector providing weights for the observations in x

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medianSamp

Aggregation function for a ordinal variable

Description

The function medianSamp chooses the level as the median or randomly between two levels.

Usage

```
medianSamp(x, weights = NULL)
```

Arguments

x ordered factor vectorweights numeric vector providing weights for the observations in x

See Also

```
Other imputation methods: hotdeck(), impPCA(), irmi(), kNN(), matchImpute(), rangerImpute(), regressionImp(), sampleCat()
```

mosaicMiss

Mosaic plot with information about missing/imputed values

Description

Create a mosaic plot with information about missing/imputed values.

```
mosaicMiss(
    x,
    delimiter = NULL,
    highlight = NULL,
    selection = c("any", "all"),
    plotvars = NULL,
    col = c("skyblue", "red", "orange"),
    labels = NULL,
    miss.labels = TRUE,
    ...
)
```

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Arguments

x	a matrix or data. frame.
delimiter	a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).
highlight	a vector giving the variables to be used for highlighting. If NULL (the default), all variables are used for highlighting.
selection	the selection method for highlighting missing/imputed values in multiple highlight variables. Possible values are "any" (highlighting of missing/imputed values in <i>any</i> of the highlight variables) and "all" (highlighting of missing/imputed values in <i>all</i> of the highlight variables).
plotvars	a vector giving the categorical variables to be plotted. If NULL (the default), all variables are plotted.
col	a vector of length three giving the colors to be used for observed, missing and imputed data. If only one color is supplied, the tiles corresponding to observed data are transparent and the supplied color is used for highlighting.
labels	a list of arguments for the labeling function vcd::labeling_border().
miss.labels	either a logical indicating whether labels should be plotted for observed and missing/imputed (highlighted) data, or a character vector giving the labels.
	additional arguments to be passed to vcd::mosaic().

Details

Mosaic plots are graphical representations of multi-way contingency tables. The frequencies of the different cells are visualized by area-proportional rectangles (tiles). Additional tiles are be used to display the frequencies of missing/imputed values. Furthermore, missing/imputed values in a certain variable or combination of variables can be highlighted in order to explore their structure.

Value

An object of class "structable" is returned invisibly.

Note

This function uses the highly flexible strucplot framework of package vcd.

Author(s)

Andreas Alfons, modifications by Bernd Prantner

60 pairs VIM

References

Meyer, D., Zeileis, A. and Hornik, K. (2006) The strucplot framework: Visualizing multi-way contingency tables with **vcd**. *Journal of Statistical Software*, **17** (3), 1–48.

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
spineMiss(), vcd::mosaic()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

pairsVIM

Scatterplot Matrices

Description

Create a scatterplot matrix.

```
pairsVIM(
    x,
    ...,
    delimiter = NULL,
    main = NULL,
    sub = NULL,
    panel = points,
    lower = panel,
    upper = panel,
    diagonal = NULL,
    labels = TRUE,
    pos.labels = NULL,
```

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```
cex.labels = NULL,
font.labels = par("font"),
layout = c("matrix", "graph"),
gap = 1
)
```

Arguments

x a matrix or data. frame.

... further arguments and graphical parameters to be passed down. par("oma")

will be set appropriately unless supplied (see graphics::par()).

delimiter a character-vector to distinguish between variables and imputation-indices for

imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

main, sub main and sub title.

panel a function $(x, y, ...\{\})$, which is used to plot the contents of each off-diagonal

panel of the display.

lower, upper separate panel functions to be used below and above the diagonal, respectively.

diagonal optional function $(x, ...\{\})$ to be applied on the diagonal panels.

labels either a logical indicating whether labels should be plotted in the diagonal pan-

els, or a character vector giving the labels.

pos.labels the vertical position of the labels in the diagonal panels.

cex.labels the character expansion factor to be used for the labels.

font.labels the font to be used for the labels.

layout a character string giving the layout of the scatterplot matrix. Possible values

are "matrix" (a matrix-like layout with the first row on top) and "graph" (a

graph-like layout with the first row at the bottom).

gap a numeric value giving the distance between the panels in margin lines.

Details

This function is the workhorse for marginmatrix() and scattmatrixMiss().

The graphical parameter oma will be set unless supplied as an argument.

A panel function should not attempt to start a new plot, since the coordinate system for each panel is set up by pairsVIM.

Note

The code is based on graphics::pairs(). Starting with version 1.4, infinite values are no longer removed before passing the x and y vectors to the panel functions.

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Author(s)

Andreas Alfons, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
marginmatrix(), scattmatrixMiss()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), parcoordMiss(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

```
data(sleep, package = "VIM")
x <- sleep[, -(8:10)]
x[,c(1,2,4,6,7)] <- log10(x[,c(1,2,4,6,7)])
pairsVIM(x)</pre>
```

parcoordMiss

Parallel coordinate plot with information about missing/imputed values

Description

Parallel coordinate plot with adjustments for missing/imputed values. Missing values in the plotted variables may be represented by a point above the corresponding coordinate axis to prevent disconnected lines. In addition, observations with missing/imputed values in selected variables may be highlighted.

```
parcoordMiss(
    x,
    delimiter = NULL,
    highlight = NULL,
    selection = c("any", "all"),
    plotvars = NULL,
    plotNA = TRUE,
    col = c("skyblue", "red", "skyblue4", "red4", "orange", "orange4"),
    alpha = NULL,
    lty = par("lty"),
```

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```
xlim = NULL,
 ylim = NULL,
 main = NULL,
  sub = NULL,
  xlab = NULL,
  ylab = NULL,
  labels = TRUE,
  xpd = NULL,
  interactive = TRUE.
)
```

Arguments

a matrix or data. frame. Χ

delimiter a character-vector to distinguish between variables and imputation-indices for

> imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

highlight a vector giving the variables to be used for highlighting. If NULL (the default),

all variables are used for highlighting.

selection the selection method for highlighting missing/imputed values in multiple high-

> light variables. Possible values are "any" (highlighting of missing/imputed values in any of the highlight variables) and "all" (highlighting of missing/imputed

values in all of the highlight variables).

plotvars a vector giving the variables to be plotted. If NULL (the default), all variables are

plotted.

plotNA a logical indicating whether missing values in the plot variables should be rep-

resented by a point above the corresponding coordinate axis to prevent discon-

nected lines.

col if plotNA is TRUE, a vector of length six giving the colors to be used for obser-

> vations with different combinations of observed and missing/imputed values in the plot variables and highlight variables (vectors of length one or two are recycled). Otherwise, a vector of length two giving the colors for non-highlighted and highlighted observations (if a single color is supplied, it is used for both).

alpha a numeric value between 0 and 1 giving the level of transparency of the colors,

or NULL. This can be used to prevent overplotting.

lty if plotNA is TRUE, a vector of length four giving the line types to be used for

> observations with different combinations of observed and missing/imputed values in the plot variables and highlight variables (vectors of length one or two are recycled). Otherwise, a vector of length two giving the line types for nonhighlighted and highlighted observations (if a single line type is supplied, it is

used for both).

xlim, ylim axis limits. 64 parcoordMiss

main, sub main and sub title.

xlab, ylab axis labels.

labels either a logical indicating whether labels should be plotted below each coordi-

nate axis, or a character vector giving the labels.

xpd a logical indicating whether the lines should be allowed to go outside the plot

region. If NULL, it defaults to TRUE unless axis limits are specified.

interactive a logical indicating whether interactive features should be enabled (see 'De-

ails').

... for parcoordMiss, further graphical parameters to be passed down (see graphics::par()).

For TKRparcoordMiss, further arguments to be passed to parcoordMiss.

Details

In parallel coordinate plots, the variables are represented by parallel axes. Each observation of the scaled data is shown as a line. Observations with missing/imputed values in selected variables may thereby be highlighted. However, plotting variables with missing values results in disconnected lines, making it impossible to trace the respective observations across the graph. As a remedy, missing values may be represented by a point above the corresponding coordinate axis, which is separated from the main plot by a small gap and a horizontal line, as determined by plotNA. Connected lines can then be drawn for all observations. Nevertheless, a caveat of this display is that it may draw attention away from the main relationships between the variables.

If interactive is TRUE, it is possible switch between this display and the standard display without the separate level for missing values by clicking in the top margin of the plot. In addition, the variables to be used for highlighting can be selected interactively. Observations with missing/imputed values in any or in all of the selected variables are highlighted (as determined by selection). A variable can be added to the selection by clicking on a coordinate axis. If a variable is already selected, clicking on its coordinate axis removes it from the selection. Clicking anywhere outside the plot region (except the top margin, if missing/imputed values exist) quits the interactive session.

Note

Some of the argument names and positions have changed with versions 1.3 and 1.4 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments colcomb and xaxlabels can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use highlight and labels instead.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

References

Wegman, E. J. (1990) Hyperdimensional data analysis using parallel coordinates. *Journal of the American Statistical Association* **85** (411), 664–675.

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

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See Also

```
pbox()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), pairsVIM(), pbox(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

```
data(chorizonDL, package = "VIM")
## for missing values
parcoordMiss(chorizonDL[,c(15,101:110)],
    plotvars=2:11, interactive = FALSE)
legend("top", col = c("skyblue", "red"), lwd = c(1,1),
    legend = c("observed in Bi", "missing in Bi"))

## for imputed values
parcoordMiss(kNN(chorizonDL[,c(15,101:110)]), delimiter = "_imp" ,
    plotvars=2:11, interactive = FALSE)
legend("top", col = c("skyblue", "orange"), lwd = c(1,1),
    legend = c("observed in Bi", "imputed in Bi"))
```

pbox

Parallel boxplots with information about missing/imputed values

Description

Boxplot of one variable of interest plus information about missing/imputed values in other variables.

```
pbox(
  Х,
 delimiter = NULL,
 pos = 1,
  selection = c("none", "any", "all"),
  col = c("skyblue", "red", "red4", "orange", "orange4"),
  numbers = TRUE,
  cex.numbers = par("cex"),
  xlim = NULL,
 ylim = NULL,
 main = NULL,
  sub = NULL,
  xlab = NULL,
 ylab = NULL,
  axes = TRUE,
  frame.plot = axes,
```

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```
labels = axes,
  interactive = TRUE,
)
```

Arguments

pos

a vector, matrix or data. frame. Х

delimiter a character-vector to distinguish between variables and imputation-indices for

> imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors

are adjusted according to the given colors for imputed variables (see col).

a numeric value giving the index of the variable of interest. Additional variables in x are used for grouping according to missingness/number of imputed

missings.

selection the selection method for grouping according to missingness/number of imputed

> missings in multiple additional variables. Possible values are "none" (grouping according to missingness/number of imputed missings in every other variable that contains missing/imputed values), "any" (grouping according to missingness/number of imputed missings in any of the additional variables) and "all" (grouping according to missingness/number of imputed missings in all of the

additional variables).

a vector of length five giving the colors to be used in the plot. The first color col

is used for the boxplots of the available data, the second/fourth are used for missing/imputed data, respectively, and the third/fifth color for the frequencies of missing/imputed values in both variables (see 'Details'). If only one color is supplied, it is used for the boxplots for missing/imputed data, whereas the boxplots for the available data are transparent. Else if two colors are supplied,

the second one is recycled.

a logical indicating whether the frequencies of missing/imputed values should numbers

be displayed (see 'Details').

cex.numbers the character expansion factor to be used for the frequencies of the missing/imputed

values.

xlim, ylim axis limits.

main, sub main and sub title.

xlab, ylab axis labels.

axes a logical indicating whether axes should be drawn on the plot.

frame.plot a logical indicating whether a box should be drawn around the plot.

labels either a logical indicating whether labels should be plotted below each box, or a

character vector giving the labels.

interactive a logical indicating whether variables can be switched interactively (see 'De-

tails').

for pbox, further arguments and graphical parameters to be passed to graphics::boxplot()

and other functions. For TKRpbox, further arguments to be passed to pbox.

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Details

This plot consists of several boxplots. First, a standard boxplot of the variable of interest is produced. Second, boxplots grouped by observed and missing/imputed values according to selection are produced for the variable of interest.

Additionally, the frequencies of the missing/imputed values can be represented by numbers. If so, the first line corresponds to the observed values of the variable of interest and their distribution in the different groups, the second line to the missing/imputed values.

If interactive=TRUE, clicking in the left margin of the plot results in switching to the previous variable and clicking in the right margin results in switching to the next variable. Clicking anywhere else on the graphics device quits the interactive session.

Value

```
a list as returned by graphics::boxplot().
```

Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments names and cex.text can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use labels and cex.numbers instead.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
parcoordMiss()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), scattJitt(), scattMiss(), scattmatrixMiss(),
spineMiss()
```

Examples

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prepare

Transformation and standardization

Description

This function is used by the VIM GUI for transformation and standardization of the data.

Usage

```
prepare(
    x,
    scaling = c("none", "classical", "MCD", "robust", "onestep"),
    transformation = c("none", "minus", "reciprocal", "logarithm", "exponential",
        "boxcox", "clr", "ilr", "alr"),
    alpha = NULL,
    powers = NULL,
    start = 0,
    alrVar
)
```

Arguments

```
a vector, matrix or data. frame.
                  the scaling to be applied to the data. Possible values are "none", "classical",
scaling
                  MCD, "robust" and "onestep".
transformation the transformation of the data. Possible values are "none", "minus", "reciprocal",
                  "logarithm", "exponential", "boxcox", "clr", "ilr" and "alr".
alpha
                  a numeric parameter controlling the size of the subset for the MCD (if scaling="MCD").
                  See robustbase::covMcd().
                  a numeric vector giving the powers to be used in the Box-Cox transformation (if
powers
                  transformation="boxcox"). If NULL, the powers are calculated with function
                  car::powerTransform().
start
                  a constant to be added prior to Box-Cox transformation (if transformation="boxcox").
                  variable to be used as denominator in the additive logratio transformation (if
alrVar
                  transformation="alr").
```

Details

Transformation:

"none": no transformation is used.

```
"logarithm": compute the logarithm (to the base 10).
```

"boxcox": apply a Box-Cox transformation. Powers may be specified or calculated with the function car::powerTransform().

Standardization:

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```
"none": no standardization is used.

"classical": apply a z-Transformation on each variable by using function scale().

"robust": apply a robustified z-Transformation by using median and MAD.
```

Value

Transformed and standardized data.

Author(s)

Matthias Templ, modifications by Andreas Alfons

See Also

```
scale(), car::powerTransform()
```

Examples

```
data(sleep, package = "VIM")
x <- sleep[, c("BodyWgt", "BrainWgt")]
prepare(x, scaling = "robust", transformation = "logarithm")</pre>
```

pulplignin

Pulp lignin content

Description

Pulp quality by lignin content remaining

Format

A data frame with 301 observations on the following 23 variables.

Details

Pulp quality is measured by the lignin content remaining in the pulp: the Kappa number. This data set is used to understand which variables in the process influence the Kappa number, and if it can be predicted accurately enough for an inferential sensor application. Variables with a number at the end have been lagged by that number of hours to line up the data.

Source

```
https://openmv.net/info/kamyr-digester
```

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References

K. Walkush and R.R. Gustafson. Application of feedforward neural networks and partial least squares regression for modelling Kappa number in a continuous Kamyr digester", Pulp and Paper Canada, 95, 1994, p T7-T13.

Examples

```
data(pulplignin)
str(pulplignin)
aggr(pulplignin)
```

rangerImpute

Random Forest Imputation

Description

Impute missing values based on a random forest model using ranger::ranger()

Usage

```
rangerImpute(
  formula,
  data,
  imp_var = TRUE,
  imp_suffix = "imp",
  ...,
  verbose = FALSE,
  median = FALSE
)
```

Arguments

formula	model formula for the imputation
data	A data. frame containing the data
imp_var	TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be created show the imputation status
<pre>imp_suffix</pre>	suffix used for TF imputation variables
	Arguments passed to ranger::ranger()
verbose	Show the number of observations used for training and evaluating the RF-Model. This parameter is also passed down to ranger::ranger() to show computation status.
median	Use the median (rather than the arithmetic mean) to average the values of individual trees for a more robust estimate.

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Value

the imputed data set.

See Also

```
Other imputation methods: hotdeck(), impPCA(), irmi(), kNN(), matchImpute(), medianSamp(), regressionImp(), sampleCat()
```

Examples

```
data(sleep)
rangerImpute(Dream+NonD~BodyWgt+BrainWgt,data=sleep)
```

regressionImp

Regression Imputation

Description

Impute missing values based on a regression model.

Usage

```
regressionImp(
  formula,
  data,
  family = "AUTO",
  robust = FALSE,
  imp_var = TRUE,
  imp_suffix = "imp",
  mod_cat = FALSE
)
```

Arguments

formula	model formula to impute one variable
data	A data.frame containing the data
family	family argument for $glm()$. "AUTO" (the default) tries to choose automatically and is the only really tested option!!!
robust	TRUE/FALSE if robust regression should be used. See details.
imp_var	TRUE/FALSE if a TRUE/FALSE variables for each imputed variable should be created show the imputation status $$
<pre>imp_suffix</pre>	suffix used for TF imputation variables
mod_cat	TRUE/FALSE if TRUE for categorical variables the level with the highest prediction probability is selected, otherwise it is sampled according to the probabilities.

rugNA

Details

```
lm() is used for family "normal" and glm() for all other families. (robust=TRUE: lmrob(),
glmrob())
```

Value

the imputed data set.

Author(s)

Alexander Kowarik

References

A. Kowarik, M. Templ (2016) Imputation with R package VIM. *Journal of Statistical Software*, 74(7), 1-16.

See Also

```
Other imputation methods: hotdeck(), impPCA(), irmi(), kNN(), matchImpute(), medianSamp(), rangerImpute(), sampleCat()
```

Examples

```
data(sleep)
sleepImp1 <- regressionImp(Dream+NonD~BodyWgt+BrainWgt,data=sleep)
sleepImp2 <- regressionImp(Sleep+Gest+Span+Dream+NonD~BodyWgt+BrainWgt,data=sleep)

data(testdata)
imp_testdata1 <- regressionImp(b1+b2~x1+x2,data=testdata$wna)
imp_testdata3 <- regressionImp(x1~x2,data=testdata$wna,robust=TRUE)</pre>
```

rugNA

Rug representation of missing/imputed values

Description

Add a rug representation of missing/imputed values in only one of the variables to scatterplots.

```
rugNA(
    x,
    y,
    ticksize = NULL,
    side = 1,
    col = "red",
```

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```
alpha = NULL,
miss = NULL,
lwd = 0.5,
...
)
```

Arguments

x, y	numeric vectors.
ticksize	the length of the ticks. Positive lengths give inward ticks.
side	an integer giving the side of the plot to draw the rug representation.
col	the color to be used for the ticks.
alpha	the alpha value (between 0 and 1).
miss	a data.frame or matrix with two columns and logical values. If NULL, x and y are searched for missing values, otherwise, the first column of miss is used to determine the imputed values in x and the second one for the imputed values in
	y.
lwd	the line width to be used for the ticks.
	further arguments to be passed to graphics::Axis().

Details

If side is 1 or 3, the rug representation consists of values available in x but missing/imputed in y. Else if side is 2 or 4, it consists of values available in y but missing/imputed in x.

Author(s)

Andreas Alfons, modifications by Bernd Prantner

Examples

```
data(tao, package = "VIM")
## for missing values
x <- tao[, "Air.Temp"]
y <- tao[, "Humidity"]
plot(x, y)
rugNA(x, y, side = 1)
rugNA(x, y, side = 2)

## for imputed values
x_imp <- kNN(tao[, c("Air.Temp","Humidity")])
x <- x_imp[, "Air.Temp"]
y <- x_imp[, "Humidity"]
miss <- x_imp[, c("Air.Temp_imp","Humidity_imp")]
plot(x, y)
rugNA(x, y, side = 1, col = "orange", miss = miss)
rugNA(x, y, side = 2, col = "orange", miss = miss)</pre>
```

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sampleCat

Random aggregation function for a factor variable

Description

The function sampleCat samples with probabilites corresponding to the occurrence of the level in the NNs.

Usage

```
sampleCat(x, weights = NULL)
```

Arguments

x factor vector

weights numeric vector providing weights for the observations in x

See Also

Other imputation methods: hotdeck(), impPCA(), irmi(), kNN(), matchImpute(), medianSamp(), rangerImpute(), regressionImp()

SBS5242

Synthetic subset of the Austrian structural business statistics data

Description

Synthetic subset of the Austrian structural business statistics (SBS) data, namely NACE code 52.42 (retail sale of clothing).

Details

The Austrian SBS data set consists of more than 320.000 enterprises. Available raw (unedited) data set: 21669 observations in 90 variables, structured according NACE revision 1.1 with 3891 missing values.

We investigate 9 variables of NACE 52.42 (retail sale of clothing).

From these confidential raw data set a non-confidential, close-to-reality, synthetic data set was generated.

Source

```
http://www.statistik.at
```

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Examples

```
data(SBS5242)
aggr(SBS5242)
```

scattJitt

Bivariate jitter plot

Description

Create a bivariate jitter plot.

Usage

```
scattJitt(
  delimiter = NULL,
  col = c("skyblue", "red", "red4", "orange", "orange4"),
  alpha = NULL,
  cex = par("cex"),
  col.line = "lightgrey",
 lty = "dashed",
  lwd = par("lwd"),
  numbers = TRUE,
  cex.numbers = par("cex"),
 main = NULL,
  sub = NULL,
 xlab = NULL,
 ylab = NULL,
  axes = TRUE,
  frame.plot = axes,
 labels = c("observed", "missing", "imputed"),
)
```

Arguments

Х

a data. frame or matrix with two columns.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

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a vector of length five giving the colors to be used in the plot. The first color will

be used for complete observations, the second/fourth color for missing/imputed values in only one variable, and the third/fifth color for missing/imputed values in both variables. If only one color is supplied, it is used for all. Else if two colors are supplied, the second one is recycled. alpha a numeric value between 0 and 1 giving the level of transparency of the colors, or NULL. This can be used to prevent overplotting. the character expansion factor for the plot characters. cex the color for the lines dividing the plot region. col.line lty the line type for the lines dividing the plot region (see graphics::par()). lwd the line width for the lines dividing the plot region. a logical indicating whether the frequencies of observed and missing/imputed numbers values should be displayed (see 'Details'). the character expansion factor to be used for the frequencies of the observed and cex.numbers missing/imputed values. main, sub main and sub title. xlab, ylab axis labels.

x and axis faucis.

axes a logical indicating whether both axes should be drawn on the plot. Use graphi-

cal parameter "xaxt" or "yaxt" to suppress just one of the axes.

frame.plot a logical indicating whether a box should be drawn around the plot.

labels a vector of length three giving the axis labels for the regions for observed, miss-

ing and imputed values (see 'Details').

... further graphical parameters to be passed down (see graphics::par()).

Details

col

The amount of observed and missing/imputed values is visualized by jittered points. Thereby the plot region is divided into up to four regions according to the existence of missing/imputed values in one or both variables. In addition, the amount of observed and missing/imputed values can be represented by a number.

Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the argument cex.text can still be supplied to . . . {} and is handled correctly. Nevertheless, it is deprecated and no longer documented. Use cex.numbers instead.

Author(s)

Matthias Templ, modifications by Andreas Alfons and Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

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See Also

```
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(), matrixplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattMiss(), scattmatrixMiss(), spineMiss()
```

Examples

```
data(tao, package = "VIM")
## for missing values
scattJitt(tao[, c("Air.Temp", "Humidity")])
## for imputed values
scattJitt(kNN(tao[, c("Air.Temp", "Humidity")]), delimiter = "_imp")
```

scattmatrixMiss

Scatterplot matrix with information about missing/imputed values

Description

Scatterplot matrix in which observations with missing/imputed values in certain variables are highlighted.

Usage

```
scattmatrixMiss(
    x,
    delimiter = NULL,
    highlight = NULL,
    selection = c("any", "all"),
    plotvars = NULL,
    col = c("skyblue", "red", "orange"),
    alpha = NULL,
    pch = c(1, 3),
    lty = par("lty"),
    diagonal = c("density", "none"),
    interactive = TRUE,
    ...
)
```

Arguments

x a matrix or data.frame.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If

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	such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).
highlight	a vector giving the variables to be used for highlighting. If NULL (the default), all variables are used for highlighting.
selection	the selection method for highlighting missing/imputed values in multiple highlight variables. Possible values are "any" (highlighting of missing/imputed values in <i>any</i> of the highlight variables) and "all" (highlighting of missing/imputed values in <i>all</i> of the highlight variables).
plotvars	a vector giving the variables to be plotted. If \ensuremath{NULL} (the default), all variables are plotted.
col	a vector of length three giving the colors to be used in the plot. The second/third color will be used for highlighting missing/imputed values.
alpha	a numeric value between 0 and 1 giving the level of transparency of the colors, or NULL. This can be used to prevent overplotting.
pch	a vector of length two giving the plot characters. The second plot character will be used for the highlighted observations.
lty	a vector of length two giving the line types for the density plots in the diagonal panels (if diagonal="density"). The second line type is used for the highlighted observations. If a single value is supplied, it is used for both non-highlighted and highlighted observations.
diagonal	a character string specifying the plot to be drawn in the diagonal panels. Possible values are "density" (density plots for non-highlighted and highlighted observations) and "none".
interactive	a logical indicating whether the variables to be used for highlighting can be selected interactively (see 'Details').
•••	for scattmatrixMiss, further arguments and graphical parameters to be passed to pairsVIM(). par("oma") will be set appropriately unless supplied (see graphics::par()). For TKRscattmatrixMiss, further arguments to be passed to scattmatrixMiss.

Details

scattmatrixMiss uses pairsVIM() with a panel function that allows highlighting of missing/imputed values.

If interactive=TRUE, the variables to be used for highlighting can be selected interactively. Observations with missing/imputed values in any or in all of the selected variables are highlighted (as determined by selection). A variable can be added to the selection by clicking in a diagonal panel. If a variable is already selected, clicking on the corresponding diagonal panel removes it from the selection. Clicking anywhere else quits the interactive session.

The graphical parameter oma will be set unless supplied as an argument.

TKRscattmatrixMiss behaves like scattmatrixMiss, but uses tkrplot to embed the plot in a Tcl/Tk window. This is useful if the number of variables is large, because scrollbars allow to move from one part of the plot to another.

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Note

Some of the argument names and positions have changed with version 1.3 due to a re-implementation and for more consistency with other plot functions in VIM. For back compatibility, the argument colcomb can still be supplied to ...{} and is handled correctly. Nevertheless, it is deprecated and no longer documented. Use highlight instead. The arguments smooth, reg.line and legend.plot are no longer used and ignored if supplied.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
pairsVIM(), marginmatrix()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(),
spineMiss()
```

Examples

```
data(sleep, package = "VIM")
## for missing values
x <- sleep[, 1:5]
x[,c(1,2,4)] <- log10(x[,c(1,2,4)])
scattmatrixMiss(x, highlight = "Dream")

## for imputed values
x_imp <- kNN(sleep[, 1:5])
x_imp[,c(1,2,4)] <- log10(x_imp[,c(1,2,4)])
scattmatrixMiss(x_imp, delimiter = "_imp", highlight = "Dream")</pre>
```

scattMiss

Scatterplot with information about missing/imputed values

Description

In addition to a standard scatterplot, lines are plotted for the missing values in one variable. If there are imputed values, they will be highlighted.

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Usage

```
scattMiss(
  Χ,
  delimiter = NULL,
  side = 1,
  col = c("skyblue", "red", "orange", "lightgrey"),
  alpha = NULL,
  lty = c("dashed", "dotted"),
  lwd = par("lwd"),
  quantiles = c(0.5, 0.975),
  inEllipse = FALSE,
  zeros = FALSE,
  xlim = NULL.
 ylim = NULL,
 main = NULL,
  sub = NULL,
  xlab = NULL,
  ylab = NULL,
  interactive = TRUE,
)
```

Arguments

Χ

a matrix or data. frame with two columns.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

side

if side=1, a rug representation and vertical lines are plotted for the missing/imputed values in the second variable; if side=2, a rug representation and horizontal lines for the missing/imputed values in the first variable.

col

a vector of length four giving the colors to be used in the plot. The first color is used for the scatterplot, the second/third color for the rug representation for missing/imputed values. The second color is also used for the lines for missing values. Imputed values will be highlighted with the third color, and the fourth color is used for the ellipses (see 'Details'). If only one color is supplied, it is used for the scatterplot, the rug representation and the lines, whereas the default color is used for the ellipses. Else if a vector of length two is supplied, the default color is used for the ellipses as well.

alpha

a numeric value between 0 and 1 giving the level of transparency of the colors, or NULL. This can be used to prevent overplotting.

1ty

a vector of length two giving the line types for the lines and ellipses. If a single value is supplied, it will be used for both.

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lwd	a vector of length two giving the line widths for the lines and ellipses. If a single value is supplied, it will be used for both.
quantiles	a vector giving the quantiles of the chi-square distribution to be used for the tolerance ellipses, or NULL to suppress plotting ellipses (see 'Details').
inEllipse	plot lines only inside the largest ellipse. Ignored if quantiles is NULL or if there are imputed values.
zeros	a logical vector of length two indicating whether the variables are semi-continuous, i.e., contain a considerable amount of zeros. If TRUE, only the non-zero observations are used for computing the tolerance ellipses. If a single logical is supplied, it is recycled. Ignored if quantiles is NULL.
xlim, ylim	axis limits.
main, sub	main and sub title.
xlab, ylab	axis labels.
interactive	a logical indicating whether the side argument can be changed interactively (see 'Details').
	further graphical parameters to be passed down (see graphics::par()).

Details

Information about missing values in one variable is included as vertical or horizontal lines, as determined by the side argument. The lines are thereby drawn at the observed x- or y-value. In case of imputed values, they will additionally be highlighted in the scatterplot. Supplementary, percentage coverage ellipses can be drawn to give a clue about the shape of the bivariate data distribution.

If interactive TRUE, clicking in the bottom margin redraws the plot with information about missing/imputed values in the first variable and clicking in the left margin redraws the plot with information about missing/imputed values in the second variable. Clicking anywhere else in the plot quits the interactive session.

Note

The argument zeros has been introduced in version 1.4. As a result, some of the argument positions have changed.

Author(s)

Andreas Alfons, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

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See Also

```
marginplot()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattmatrixMiss(),
spineMiss()
```

Examples

```
data(tao, package = "VIM")
## for missing values
scattMiss(tao[,c("Air.Temp", "Humidity")])
## for imputed values
scattMiss(kNN(tao[,c("Air.Temp", "Humidity")]), delimiter = "_imp")
```

sleep

Mammal sleep data

Description

Sleep data with missing values.

Format

A data frame with 62 observations on the following 10 variables.

BodyWgt a numeric vector

BrainWgt a numeric vector

NonD a numeric vector

Dream a numeric vector

Sleep a numeric vector

Span a numeric vector

Gest a numeric vector

Pred a numeric vector

Exp a numeric vector

Danger a numeric vector

Source

Allison, T. and Chichetti, D. (1976) Sleep in mammals: ecological and constitutional correlates. *Science* **194** (**4266**), 732–734.

The data set was imported from GGobi.

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Examples

```
data(sleep, package = "VIM")
summary(sleep)
aggr(sleep)
```

spineMiss

Spineplot with information about missing/imputed values

Description

Spineplot or spinogram with highlighting of missing/imputed values in other variables by splitting each cell into two parts. Additionally, information about missing/imputed values in the variable of interest is shown on the right hand side.

Usage

```
spineMiss(
  Х,
 delimiter = NULL,
  pos = 1,
  selection = c("any", "all"),
 breaks = "Sturges",
  right = TRUE,
  col = c("skyblue", "red", "skyblue4", "red4", "orange", "orange4"),
 border = NULL,
 main = NULL,
  sub = NULL,
 xlab = NULL,
 ylab = NULL,
  axes = TRUE,
  labels = axes,
 only.miss = TRUE,
 miss.labels = axes,
  interactive = TRUE,
)
```

Arguments

...

a vector, matrix or data. frame.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

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pos	a numeric value giving the index of the variable of interest. Additional variables in x are used for highlighting.
selection	the selection method for highlighting missing/imputed values in multiple additional variables. Possible values are "any" (highlighting of missing/imputed values in <i>any</i> of the additional variables) and "all" (highlighting of missing/imputed values in <i>all</i> of the additional variables).
breaks	if the variable of interest is numeric, breaks controls the breakpoints (see <pre>graphics::hist()</pre> for possible values).
right	logical; if TRUE and the variable of interest is numeric, the spinogram cells are right-closed (left-open) intervals.
col	a vector of length six giving the colors to be used. If only one color is supplied, the bars are transparent and the supplied color is used for highlighting missing/imputed values. Else if two colors are supplied, they are recycled.
border	the color to be used for the border of the cells. Use border=NA to omit borders.
main, sub	main and sub title.
xlab, ylab	axis labels.
axes	a logical indicating whether axes should be drawn on the plot.
labels	if the variable of interest is categorical, either a logical indicating whether labels should be plotted below each cell, or a character vector giving the labels. This is ignored if the variable of interest is numeric.
only.miss	logical; if TRUE, the missing/imputed values in the variable of interest are also visualized by a cell in the spineplot or spinogram. Otherwise, a small spineplot is drawn on the right hand side (see 'Details').
miss.labels	either a logical indicating whether label(s) should be plotted below the cell(s) on the right hand side, or a character string or vector giving the label(s) (see 'Details').
interactive	a logical indicating whether the variables can be switched interactively (see 'Details').
	further graphical parameters to be passed to graphics::title() and graphics::axis().

Details

A spineplot is created if the variable of interest is categorial and a spinogram if it is numerical. The horizontal axis is scaled according to relative frequencies of the categories/classes. If more than one variable is supplied, the cells are split according to missingness/number of imputed values in the additional variables. Thus the proportion of highlighted observations in each category/class is displayed on the vertical axis. Since the height of each cell corresponds to the proportion of highlighted observations, it is now possible to compare the proportions of missing/imputed values among the different categories/classes.

If only.miss=TRUE, the missing/imputed values in the variable of interest are also visualized by a cell in the spine plot or spinogram. If additional variables are supplied, this cell is again split into two parts according to missingness/number if imputed values in the additional variables.

Otherwise, a small spineplot that visualizes missing/imputed values in the variable of interest is drawn on the right hand side. The first cell corresponds to observed values and the second cell to

spineMiss 85

missing/imputed values. Each of the two cells is again split into two parts according to missingness/number of imputed values in the additional variables. Note that this display does not make sense if only one variable is supplied, therefore only.miss is ignored in that case.

If interactive=TRUE, clicking in the left margin of the plot results in switching to the previous variable and clicking in the right margin results in switching to the next variable. Clicking anywhere else on the graphics device quits the interactive session.

Value

a table containing the frequencies corresponding to the cells.

Note

Some of the argument names and positions have changed with version 1.3 due to extended functionality and for more consistency with other plot functions in VIM. For back compatibility, the arguments xaxlabels and missaxlabels can still be supplied to ...{} and are handled correctly. Nevertheless, they are deprecated and no longer documented. Use labels and miss.labels instead

The code is based on the function graphics::spineplot() by Achim Zeileis.

Author(s)

Andreas Alfons, Matthias Templ, modifications by Bernd Prantner

References

M. Templ, A. Alfons, P. Filzmoser (2012) Exploring incomplete data using visualization tools. *Journal of Advances in Data Analysis and Classification*, Online first. DOI: 10.1007/s11634-011-0102-y.

See Also

```
histMiss(), barMiss(), mosaicMiss()
Other plotting functions: aggr(), barMiss(), histMiss(), marginmatrix(), marginplot(),
matrixplot(), mosaicMiss(), pairsVIM(), parcoordMiss(), pbox(), scattJitt(), scattMiss(),
scattmatrixMiss()
```

Examples

```
data(tao, package = "VIM")
data(sleep, package = "VIM")
## for missing values
spineMiss(tao[, c("Air.Temp", "Humidity")])
spineMiss(sleep[, c("Exp", "Sleep")])
## for imputed values
spineMiss(kNN(tao[, c("Air.Temp", "Humidity")]), delimiter = "_imp")
spineMiss(kNN(sleep[, c("Exp", "Sleep")]), delimiter = "_imp")
```

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tableMiss

create table with highlighted missings/imputations

Description

Create a reactable table that highlights missing values and imputed values with the same colors as histMiss()

Usage

```
tableMiss(x, delimiter = "_imp")
```

Arguments

x a vector, matrix or data.frame.

delimiter

a character-vector to distinguish between variables and imputation-indices for imputed variables (therefore, x needs to have colnames()). If given, it is used to determine the corresponding imputation-index for any imputed variable (a logical-vector indicating which values of the variable have been imputed). If such imputation-indices are found, they are used for highlighting and the colors are adjusted according to the given colors for imputed variables (see col).

Examples

```
data(tao)
x_IMPUTED <- kNN(tao[, c("Air.Temp", "Humidity")])
tableMiss(x_IMPUTED[105:114, ])
x_IMPUTED[106, 2] <- NA
x_IMPUTED[105, 1] <- NA
x_IMPUTED[107, "Humidity_imp"] <- TRUE
tableMiss(x_IMPUTED[105:114, ])</pre>
```

tao

Tropical Atmosphere Ocean (TAO) project data

Description

A small subsample of the Tropical Atmosphere Ocean (TAO) project data, derived from the GG0BI project.

testdata 87

Format

Year a numeric vector

A data frame with 736 observations on the following 8 variables.

Latitude a numeric vector

Longitude a numeric vector

Sea.Surface.Temp a numeric vector

Air.Temp a numeric vector

Humidity a numeric vector

UWind zonal wind, i.e. latitude-parallel wind

VWind meridional wind, i.e. longitude-parallel wind

Details

All cases recorded for five locations and two time periods.

Source

```
http://www.pmel.noaa.gov/tao/
```

Examples

```
data(tao, package = "VIM")
summary(tao)
aggr(tao)
```

testdata

Simulated data set for testing purpose

Description

2 numeric, 2 binary, 2 nominal and 2 mixed (semi-continous) variables

Format

The format is: List of 4

- \$wna: a data.frame with 500 obs. of 8 variables:
 - x1: numeric 10.87 9.53 7.83 8.53 8.67 ...
 - x2: numeric 10.9 9.32 7.68 8.2 8.41
 - c1: Factor w/ 4 levels "a", "b", "c", "d": 3 2 2 1 2 2 1 3 3 2 ...
 - c2: Factor w/ 4 levels "a", "b", "c", "d": 2 3 2 2 2 2 2 4 2 2 ...
 - b1: Factor w/ 2 levels "0","1": 2 2 1 2 1 2 1 2 1 1 ...
 - b2: Factor w/ 2 levels "0","1": 2 2 1 1 1 1 1 2 2 2 ...

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```
- m1: numeric 0 8.29 9.08 0 0 ...
```

- m2: numeric 10.66 9.39 7.8 8.11 7.33 ...
- \$wona: a 'data.frame" with 500 obs. of 8 variables:
 - x1: numeric 10.87 9.53 7.83 8.53 8.67 ...
 - x2: numeric 10.9 9.32 7.68 8.2 8.41 ...
 - c1: Factor w/ 4 levels "a", "b", "c", "d": 3 2 2 1 2 2 1 3 3 2 ...
 - c2: Factor w/ 4 levels "a", "b", "c", "d": 2 3 2 2 2 2 2 4 2 2 ...
 - b1: Factor w/ 2 levels "0","1": 2 2 1 2 1 2 1 2 1 1 ...
 - b2: Factor w/ 2 levels "0","1": 2 2 1 1 1 1 1 2 2 2 ...
 - m1: numeric 0 8.29 9.08 0 0 ...
 - m2: numeric 10.66 9.39 7.8 8.11 7.33 ...
- \$mixed: c("m1", "m2")
- \$outlierInd: 'NULL"

Examples

data(testdata)

toydataMiss

Simulated toy data set for examples

Description

A 2-dimensional data set with additional information.

Format

data frame with 100 observations and 12 variables. The first two variables represent the fully observed data.

Examples

data(toydataMiss)

wine 89

wine

Wine tasting and price

Description

Wine reviews from France, Switzerland, Austria and Germany.

Format

A data frame with 9627 observations on the following 9 variables.

country country of origin

points the number of points WineEnthusiast rated the wine on a scale of 1-100 (though they say they only post reviews for wines that score >=80)

price the cost for a bottle of the wine

province the province or state that the wine is from

taster_name name of the person who tasted and reviewed the wine

taster_twitter_handle Twitter handle for the person who tasted ane reviewed the wine

variety the type of grapes used to make the wine (ie pinot noir)

winery the winery that made the wine

variety_main broader category as variety

Details

The data was scraped from WineEnthusiast during the week of Nov 22th, 2017. The code for the scraper can be found at https://github.com/zackthoutt/wine-deep-learning This data set is slightly modified, i.e. only four countries are selected and broader categories on the variety have been added.

Source

https://www.kaggle.com/zynicide/wine-reviews

Examples

data(wine)
str(wine)
aggr(wine)

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