

Package ‘opticskxi’

December 9, 2024

Title OPTICS K-Xi Density-Based Clustering

Version 1.1.0

Description Density-based clustering methods are well adapted to the clustering of high-dimensional data and enable the discovery of core groups of various shapes despite large amounts of noise. This package provides a novel density-based cluster extraction method, OPTICS k-Xi, and a framework to compare k-Xi models using distance-based metrics to investigate datasets with unknown number of clusters. The vignette first introduces density-based algorithms with simulated datasets, then presents and evaluates the k-Xi cluster extraction method. Finally, the models comparison framework is described and experimented on 2 genetic datasets to identify groups and their discriminating features. The k-Xi algorithm is a novel OPTICS cluster extraction method that specifies directly the number of clusters and does not require fine-tuning of the steepness parameter as the OPTICS Xi method. Combined with a framework that compares models with varying parameters, the OPTICS k-Xi method can identify groups in noisy datasets with unknown number of clusters. Results on summarized genetic data of 1,200 patients are in Charlon T. (2019) <[doi:10.13097/archive-ouverte/unige:161795](https://doi.org/10.13097/archive-ouverte/unige:161795)>.

Imports ggplot2, magrittr, rlang

Depends R (>= 3.5.0)

Suggests amap, dbscan, cowplot, fastICA, fpc, ggrepel, grid, grDevices, gtable, knitr, parallel, plyr, reshape2, stats, testthat, text2vec, utils

VignetteBuilder knitr

License GPL-3

Encoding UTF-8

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contingency_table	<i>Contingency table</i>
-------------------	--------------------------

Description

Include NAs and add totals to table.

Usage

```
contingency_table(...)
```

Arguments

... Passed to table

Value

Table object

crohn	<i>Crohn's disease data</i>
-------	-----------------------------

Description

The data set consist of 103 common (>5% minor allele frequency) SNPs genotyped in 129 trios from an European-derived population. These SNPs are in a 500-kb region on human chromosome 5q31 implicated as containing a genetic risk factor for Crohn disease.

Imported from the gap R package.

An example use of the data is with the following paper, Kelly M. Burkett, Celia M. T. Greenwood, BradMcNeney, Jinko Graham. Gene genealogies for genetic association mapping, with application to Crohn's disease. Fron Genet 2013, 4(260) doi: 10.3389/fgene.2013.00260

Usage

```
data(crohn)
```

Format

A data frame containing 387 rows and 212 columns

Source

MJ Daly, JD Rioux, SF Schaffner, TJ Hudson, ES Lander (2001) High-resolution haplotype structure in the human genome Nature Genetics 29:229-232

ensemble_metrics	<i>Compute ensemble metrics</i>
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Description

Use models' rankings over several metrics to select best model.

Usage

```
ensemble_metrics(  
  n_top = 0,  
  df_params,  
  metrics = NULL,  
  metrics_exclude = NULL,  
  n_models = 10  
)
```

Arguments

n_top	Threshold of number of models to rank
df_params	Output of opticksxi_pipeline
metrics	Names of metrics to use. Any of those computed by opticksxi_pipeline, e.g. 'sindex', 'ch', 'dunn', 'dunn2', 'widestgap', 'entropy' etc. NULL for all (8).
metrics_exclude	Names of metrics to exclude. Typically used with metrics = NULL. E.g. 'entropy'.
n_models	Number of best models to return

Value

List of metrics matrix and df_params subsetted to best models

ensemble_models *Select models based on ensemble metrics*

Description

Typically we will call ensemble_metrics with varying numbers of ranks to consider and this function will sum up the ranks from those calls.

Usage

```
ensemble_models(l_ensemble_metrics, n_models = 4)
```

Arguments

l_ensemble_metrics	Output of function ensemble_metrics
n_models	Number of best models to return

Value

List of parameters of best models

fortify_dimred	<i>Fortify a dimension reduction object</i>
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Description

Fortify a dimension reduction object

Usage

```
fortify_dimred(  
  m_dimred,  
  m_vars = NULL,  
  v_variance = NULL,  
  sup_vars = NULL,  
  var_digits = 1  
)
```

Arguments

m_dimred	Projection matrix
m_vars	Rotation matrix (optional)
v_variance	Explained variance (optional)
sup_vars	Optional supplementary variables
var_digits	Explained variance percent digits

Value

Data frame

See Also

[fortify_pca](#), [fortify_ica](#)

Examples

```
pca <- prcomp(iris[-5])  
df_pca <- fortify_dimred(pca$x)
```

fortify_ica	<i>Get and fortify ICA</i>
-------------	----------------------------

Description

Get and fortify ICA

Usage

```
fortify_ica(m_data, ..., sup_vars = NULL)
```

Arguments

m_data	Input matrix
...	Passed to fastICA::fastICA
sup_vars	Optional supplementary variables

Value

Fortified dimension reduction

See Also

[fortify_dimred](#), [fortify_pca](#)

Examples

```
df_ica <- fortify_ica(iris[-5], n.comp = 2)
```

fortify_pca	<i>Get and fortify PCA</i>
-------------	----------------------------

Description

Get and fortify PCA

Usage

```
fortify_pca(m_data, ..., sup_vars = NULL)
```

Arguments

m_data	Input matrix
...	Passed to stats::prcomp
sup_vars	Optional supplementary variables

Value

Fortified dimension reduction

See Also

[fortify_dimred](#), [fortify_ica](#)

Examples

```
df_pca <- fortify_pca(iris[-5])  
df_pca <- fortify_pca(iris[-5], sup_vars = iris[5])
```

get_best_kxi	<i>Get best k-Xi model</i>
--------------	----------------------------

Description

Select k-Xi clustering model based on a metric and a rank

Usage

```
get_best_kxi(df_kxi, metric = "avg.silwidth", rank = 1)
```

Arguments

- df_kxi Data frame returned by `opticsxi_pipeline`
- metric Metric to choose best model
- rank Rank(s) of model to choose, ordered by decreasing metric

Value

df_kxi row with specified metric and rank, simplified to a list if only one rank selected

See Also

[opticskxi_pipeline](#)

ggpairs

Plot multiple axes of a data frame or a fortified dimension reduction.

Description

Plot multiple axes of a data frame or a fortified dimension reduction.

Usage

```
ggpairs(
  df_data,
  group = NULL,
  axes = 1:2,
  variables = FALSE,
  n_vars = 0,
  ellipses = FALSE,
  ...,
  title = NULL,
  colors = if (!is.null(group)) nice_palette(df_data[[group]])
)
```

Arguments

df_data	Data frame
group	Column name of the grouping of observations
axes	Axes to plot. If more than 2, plots all pair combinations
variables	Logical, plot variable contributions of the dimension reduction to the selected axes, only for 2 axes
n_vars	Maximum number of variable contributions to plot. By default 0, for all variables.
ellipses	Logical, plot ellipses of groups
...	Passed to ggplot2 stat_ellipse if ellipses are requested
title	String to add as title, default NULL
colors	Vector of colors for each group

Value

ggmatrix

See Also

[fortify_pca](#), [fortify_ica](#)

Examples

```
df_pca <- fortify_pca(iris[-5])
ggpairs(df_pca)
df_pca <- fortify_pca(iris[-5], sup_vars = iris[5])
ggpairs(df_pca, group = 'Species', ellipses = TRUE, variables = TRUE)
```

ggplot_kxi_metrics *Ggplot OPTICS k-Xi metrics*

Description

Plot metrics of a kxi_pipeline output

Usage

```
ggplot_kxi_metrics(df_kxi, metric = c("avg.silwidth", "bw.ratio"), n = 8)
```

Arguments

df_kxi	Data frame returned by opticskxi_pipeline
metric	Vector of metrics to display from the df_kxi object
n	Number of best models for the first metric to display

Value

ggplot

See Also

[opticskxi_pipeline](#)

ggplot_optics *Ggplot optics*

Description

Plot OPTICS reachability plot.

Usage

```
ggplot_optics(
  optics_obj,
  groups = NULL,
  colors = if (!is.null(groups)) nice_palette(groups),
  segment_size = 300/nrow(df_optics)
)
```

Arguments

optics_obj	dbscan::optics object
groups	Optional vector defining groups of OPTICS observations
colors	If groups specified, vector of colors for each group
segment_size	Size for geom_segment

Value

ggplot

See Also

[opticskxi](#)

Examples

```
data('multishapes')
optics_obj <- dbscan::optics(multishapes[1:2])
ggplot_optics(optics_obj)
ggplot_optics(optics_obj,
  groups = opticskxi(optics_obj, n_xi = 5, pts = 30))
```

`gtable_kxi_profiles` *Gtable OPTICS k-Xi distance profiles*

Description

Plot OPTICS distance profiles of k-Xi clustering models

Usage

```
gtable_kxi_profiles(df_kxi, metric = "avg.silwidth", rank = 1:4, ...)
```

Arguments

df_kxi	Data frame returned by opticskxi_pipeline
metric	Metric to choose best clustering model
rank	Ranks of models to plot, ordered by decreasing model metric
...	Passed to ggplot_kxi_profile

See Also

[opticskxi_pipeline](#)

hla	<i>The HLA data</i>
-----	---------------------

Description

This data set contains HLA markers DRB, DQA, DQB and phenotypes of 271 Schizophrenia patients ($y=1$) and controls ($y=0$). Genotypes for 3 HLA loci have prefixes name (e.g., "DQB") and a suffix for each of two alleles (".a1" and ".a2").

Imported from the gap package.

Usage

```
data(hla)
```

Format

A data frame containing 271 rows and 8 columns

Source

Dr Padraig Wright of Pfizer

multishapes	<i>A dataset containing clusters of multiple shapes</i>
-------------	---

Description

Data containing clusters of any shapes. Useful for comparing density-based clustering (DBSCAN) and standard partitioning methods such as k-means clustering. Imported from the factoextra package.

Usage

```
data("multishapes")
```

Format

A data frame with 1100 observations on the following 3 variables.

x a numeric vector containing the x coordinates of observations

y a numeric vector containing the y coordinates of observations

shape a numeric vector corresponding to the cluster number of each observations.

Details

The dataset contains 5 clusters and some outliers/noises.

Examples

```
data('multishapes')
plot(multishapes[, 1], multishapes[, 2],
     col = multishapes[, 3], pch = 19, cex = 0.8)
```

m_psychwords

A dataset containing words by embeddings matrix

Description

Data containing Glove embeddings of psychological related words, useful for demonstrating the use of the modified opticksxi pipeline psychkxi.

Usage

```
data("m_psychwords")
```

Format

A matrix with 799 words in rows and 100 embedding dimensions in columns.

Details

The dataset contains 2 main hierarchical clusters and each has subclusters.

nice_palette

Nice palette

Description

Color palette

Usage

```
nice_palette(groups, rainbow = FALSE)
```

Arguments

groups Vector, each unique value will get a color
rainbow If TRUE, rainbow-like colors, else differentiate successive values

Value

Vector of colors

 opticskxi

OPTICS k-Xi clustering algorithm

Description

For each largest distance differences on the OPTICS profile, consecutive observations left and right on the OPTICS profile (i.e. lower and higher OPTICS id) will be assigned to 2 different clusters if their distance is below the distance of the edge point. If above, observations are NA. The pts parameter defines a minimum number of observations to form a valley (i.e. cluster). If the number of observations in one valley is smaller than pts, observations are set to NA.

Usage

```
opticskxi(
  optics_obj,
  n_xi,
  pts = optics_obj$minPts,
  max_loop = 50,
  verbose = FALSE
)
```

Arguments

optics_obj	Data frame returned by optics
n_xi	Number of clusters to define
pts	Minimum number of points per clusters
max_loop	Maximum iterations to find n_xi clusters
verbose	Print the ids of the largest difference considered and cluster information if they define one

Value

Vector of clusters

See Also

[opticskxi_pipeline](#), [ggplot_optics](#)

Examples

```
data('multishapes')
optics_shapes <- dbscan::optics(multishapes[1:2])
kxi_shapes <- opticskxi(optics_shapes, n_xi = 5, pts = 30)
ggplot_optics(optics_shapes, groups = kxi_shapes)
ggpairs(cbind(multishapes[1:2], kXi = kxi_shapes), group = 'kXi')
```

opticskxi_pipeline *OPTICS k-Xi models comparison pipeline*

Description

Computes OPTICS k-Xi models based on a parameter grid, binds results in a data frame, and computes distance based metrics for each model.

Usage

```
opticskxi_pipeline(
  m_data,
  df_params = expand.grid(n_xi = 1:10, pts = c(20, 30, 40), dist = c("euclidean",
    "abscorrelation"), dim_red = c("identity", "PCA", "ICA"), n_dimred_comp = c(5, 10,
    20)),
  n_cores = 1
)
```

Arguments

m_data	Data matrix
df_params	Parameter grid for the OPTICS k-Xi function call and optional dimension reduction. Required columns: n_xi, pts, dist. Optonal columns: dim_red, n_dim_red.
n_cores	Number of cores

Value

Input parameter data frame with with results binded in columns optics, clusters and metrics.

See Also

[get_best_kxi](#), [ggplot_kxi_metrics](#), [gtable_kxi_profiles](#)

Examples

```
data('hla')
m_hla <- hla[-c(1:2)] %>% scale
df_params_hla <- expand.grid(n_xi = 3:5, pts = c(20, 30),
  dist = c('manhattan', 'euclidean'))
df_kxi_hla <- opticskxi_pipeline(m_hla, df_params_hla)
ggplot_kxi_metrics(df_kxi_hla, n = 8)
gtable_kxi_profiles(df_kxi_hla) %>% plot

best_kxi_hla <- get_best_kxi(df_kxi_hla, rank = 2)
clusters_hla <- best_kxi_hla$clusters
fortify_pca(m_hla, sup_vars = data.frame(Clusters = clusters_hla)) %>%
  ggpairs('Clusters', ellipses = TRUE, variables = TRUE)
```

print_table	<i>Print table</i>
-------------	--------------------

Description

Print knitr::kable latex table with legend at bottom.

Usage

```
print_table(table_obj, label)
```

Arguments

table_obj	Table object
label	Latex label

Value

None

psych_kxi_ensemble_models	<i>Example pipeline for ensemble models</i>
---------------------------	---

Description

Example pipeline for ensemble models on mental health related natural language processing

Usage

```
psych_kxi_ensemble_models(  
  m_data,  
  ...,  
  n_models = 4,  
  metrics = NULL,  
  metrics_exclude = NULL,  
  model_subsample = c(0.1, 0.2, 0.5),  
  n_models_subsample = 10  
)
```

Arguments

m_data	Data matrix Data frame returned by optics
...	Passed to function psych_kxi_pipeline
n_models	Number of best models to return
metrics	Names of metrics to use. Any of those computed by opticksxi_pipeline, e.g. 'sindex', 'ch', 'dunn', 'dunn2', 'widestgap', 'entropy' etc. NULL for all (8).
metrics_exclude	Names of metrics to exclude. Typically used with metrics = NULL. E.g. 'entropy'.
model_subsample	Ratios of best models to consider.
n_models_subsample	Number of best models when subsampling.

Value

Input parameter data frame with with results binded in columns optics, clusters and metrics. Sub-setted to best models according to ensemble metrics.

Examples

```
data('m_psychwords')
m_psychwords = m_psychwords[1:200, 1:20]

df_params = expand.grid(n_xi = 4:5, pts = c(5, 10), dist = 'cosine',
                      dim_red = 'ICA', n_dimred_comp = 5)

df_kxi = psych_kxi_ensemble_models(m_psychwords, df_params,
                                  n_min_clusters = 2,
                                  n_models = 4,
                                  metrics = c('avg.silwidth', 'dunn'),
                                  model_subsample = c(0.4, 0.6),
                                  n_models_subsample = 4)
```

psych_kxi_pipeline *Example pipeline for mental health natural language processing*

Description

Removes too large clusters and models with less than a minimum number of clusters.

Usage

```
psych_kxi_pipeline(
  m_data,
  df_params = expand.grid(n_xi = 8:15, pts = c(15, 20, 25, 30), dist = "cosine", dim_red
    = "ICA", n_dimred_comp = c(10, 15, 20, 25)),
  max_size_ratio = 0.15,
  n_min_clusters = 5,
  n_cores = 1
)
```

Arguments

m_data	Data matrix
df_params	Parameter grid for the OPTICS k-Xi function call and optional dimension reduction. Required columns: n_xi, pts, dist. Optional columns: dim_red, n_dim_red.
max_size_ratio	Maximum size ratio of clusters
n_min_clusters	Minimum number of clusters
n_cores	Number of cores

Value

Input parameter data frame with with results binded in columns optics, clusters and metrics.

residuals_table	<i>Residuals table</i>
-----------------	------------------------

Description

Bind contingency table and Pearson Chi-squared residuals.

Usage

```
residuals_table(...)
```

Arguments

...	Passed to contingency_table and chisq.test
-----	--

Value

Matrix

`%<>%` *Magrittr pipe-assign operator*

Description

Magrittr pipe-assign operator

`%%$%` *Magrittr pipe-with operator*

Description

Magrittr pipe-with operator

`%>%` *Magrittr pipe operator*

Description

Magrittr pipe operator

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