Package 'HVT'

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Credits to Mu Sigma for their continuous support throughout the development of the package.

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displayTable

Table for displaying summary

Description

This is the main function for displaying summary from model training and scoring

Usage

```
displayTable(
  data,
  columnName,
  value,
  tableType = "summary",
  scroll = TRUE,
  limit = 100
)
```

Arguments

data List. A listed object from trainHVT or scoreHVT columnName Character. Name of the column that needs highlighting. value Numeric. The value above will be highlighted in red or green.

edaPlots 3

tableType Character. Type of table to generate ('summary', 'compression')

scroll Logical. A value to have a scroll or not in the table.

limit Numeric. A value to indicate how many rows to display. Applicable for sum-

mary tableType.

Value

A consolidated table of results

Author(s)

Vishwavani@mu-sigma.com>

See Also

trainHVT

Examples

edaPlots

plots for data analysis

Description

This is the main function that provides exploratory data analysis plots

Usage

```
edaPlots(df, time_series = FALSE, time_column)
```

Arguments

df Dataframe. A data frame object.

time_series Logical. A value to indicate whether the dataset is time series or not.

time_column Character. The name of the time column in the data frame.

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Value

Five objects which include time series plots, data distribution plots, box plots, correlation plot and a descriptive statistics table.

Author(s)

Vishwavani «vishwavani@mu-sigma.com»

Examples

getCellId

Cell ID

Description

Function to generate cell ID based on 1D sammons projection

Usage

```
getCellId(hvt.results, seed = 123)
```

Arguments

hvt.results List. A list of hvt.results obtained from the trainHVT function.

seed Numeric. Random Seed

Details

To generate cell id for the multivariate data, the data is being projected from n-dimensions to 1-dimension and the cell id is being assigned by ordering these values and finding the corresponding indexes. The output Cell id gets appended to the HVT model.

Value

Object containing Cell.ID mappings for the given hvt.results list.

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>

getOptimalCentroids 5

getOptimalCentroids

Description

Get Optimal Centroids

Usage

```
getOptimalCentroids(
    x,
    iter.max,
    algorithm,
    n_cells,
    seed = 100,
    function_to_calculate_distance_metric,
    function_to_calculate_error_metric = c("mean", "max"),
    quant.err,
    distance_metric = "L1_Norm",
    quant_method = c("kmeans", "kmedoids"),
    ...
)
```

Arguments

Data Frame. A dataframe of multivariate data. Each row corresponds to an observation, and each column corresponds to a variable. Missing values are not accepted.
 String. The type of algorithm used for quantization. Available algorithms are

String. The type of algorithm used for quantization. Available algorithms are Hartigan and Wong, "Lloyd", "Forgy", "MacQueen". (default is "Hartigan-Wong")

n_cells Numeric. Indicating the number of nodes per hierarchy.

seed Numeric. Random Seed. function_to_calculate_distance_metric

Function. The function is to find 'L1_Norm" or "L2_Norm" distances. L1_Norm is selected by default.

function_to_calculate_error_metric

Character. The error metric can be "mean" or "max". mean is selected by default

quant.err Numeric. The quantization error for the algorithm.

distance_metric

Character. The distance metric to calculate inter point distance. It can be 'L1_Norm" or "L2_Norm". L1_Norm is selected by default.

quant_method Character. The quant_method can be "kmeans" or "kmedoids". kmeans is se-

lected by default

Details

The raw data is first scaled and this scaled data is supplied as input to the vector quantization algorithm. Vector quantization technique uses a parameter called quantization error. This parameter acts as a threshold and determines the number of levels in the hierarchy. It means that, if there are 'n' number of levels in the hierarchy, then all the clusters formed till this level will have quantization error equal or greater than the threshold quantization error. The user can define the number of clusters in the first level of hierarchy and then each cluster in first level is sub-divided into the same number of clusters as there are in the first level. This process continues and each group is divided into smaller clusters as long as the threshold quantization error is met. The output of this technique will be hierarchically arranged vector quantized data.

Value

Values List. A list showing observations assigned to a cluster.	values	List. A list showing observations assigned to a cluster.
---	--------	--

maxQE List. A list corresponding to maximum QE values for each cell.

meanQE List. A list corresponding to mean QE values for each cell.

centers List. A list of quantization error for all levels and nodes.

nsize List. A list corresponding to number of observations in respective groups.

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>, Sangeet Moy Das <sangeet.das@mu-sigma.com>

```
getTransitionProbability
```

Creating Transition Probabilities list

Description

This is the main function to create transition probabilities list. The transition probability table quantifies the likelihood of transitioning from one state to another. States: The table includes the current states and the possible next states. Probabilities: For each current state, it lists the probability of transitioning to each of the next possible states.

Usage

```
getTransitionProbability(df, cellid_column, time_column)
```

Arguments

df Data frame. The input data frame should contain two columns, cell ID from

scoreHVT function and time stamp of that dataset.

cellid_column Character. Name of the column containing cell IDs.
time_column Character. Name of the column containing time stamps.

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Value

Prints and stores a nested list of data frames with transition probabilities.

Author(s)

PonAnuReka Seenivasan <ponanureka.s@mu-sigma.com>

See Also

```
trainHVT
scoreHVT
```

Examples

hvq

hvq

Description

Hierarchical Vector Quantization

Usage

```
hvq(
    x,
    min_compression_perc = NA,
    n_cells = NA,
    depth = 3,
    quant.err = 10,
    seed = 300,
    algorithm = "Hartigan-Wong",
    distance_metric = c("L1_Norm", "L2_Norm"),
    error_metric = c("mean", "max"),
    quant_method = c("kmeans", "kmedoids")
)
```

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Arguments

x Data Frame. A dataframe of multivariate data. Each row corresponds to an

observation, and each column corresponds to a variable. Missing values are not

accepted.

min_compression_perc

Numeric. An integer indicating the minimum percent compression rate to be

achieved for the dataset

n_cells Numeric. Indicating the number of nodes per hierarchy.

depth Numeric. Indicating the hierarchy depth (or) the depth of the tree (1 = no hierarchy depth)

archy, 2 = 2 levels, etc..)

quant.err Numeric. The quantization error for the algorithm.

seed Numeric. Random Seed.

algorithm String. The type of algorithm used for quantization. Available algorithms are

Hartigan and Wong, "Lloyd", "Forgy", "MacQueen". (default is "Hartigan-

Wong")

distance_metric

character. The distance metric can be 'L1_Norm" or "L2_Norm". L1_Norm is

selected by default.

error_metric character. The error metric can be "mean" or "max". mean is selected by default

quant_method character. The quant_method can be "kmeans" or "kmedoids". kmeans is se-

lected by default

Details

The raw data is first scaled and this scaled data is supplied as input to the vector quantization algorithm. Vector quantization technique uses a parameter called quantization error. This parameter acts as a threshold and determines the number of levels in the hierarchy. It means that, if there are 'n' number of levels in the hierarchy, then all the clusters formed till this level will have quantization error equal or greater than the threshold quantization error. The user can define the number of clusters in the first level of hierarchy and then each cluster in first level is sub-divided into the same number of clusters as there are in the first level. This process continues and each group is divided into smaller clusters as long as the threshold quantization error is met. The output of this technique will be hierarchically arranged vector quantized data.

Value

clusters List. A list showing each ID assigned to a cluster.

nodes.clust List. A list corresponding to nodes' details.

idnodes List. A list of ID and segments similar to nodes.clust with additional columns

for nodes ID.

error guant List. A list of quantization error for all levels and nodes.

plt.clust List. A list of logical values indicating if the quantization error was met.

summary. Summary. Output table with summary.

madPlot 9

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>, Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also

```
plotHVT
```

Examples

```
data("EuStockMarkets")
dataset <- data.frame(date = as.numeric(time(EuStockMarkets)),
DAX = EuStockMarkets[, "DAX"],
SMI = EuStockMarkets[, "SMI"],
CAC = EuStockMarkets[, "CAC"],
FTSE = EuStockMarkets[, "FTSE"])
dataset_hvt <- dataset[,-c(1)]
hvqOutput = hvq(dataset_hvt, n_cells = 5, depth = 2, quant.err = 0.2,
distance_metric='L1_Norm',error_metric='mean',quant_method="kmeans")</pre>
```

madPlot

Mean Absolute Deviation Plot

Description

Function to create Mean Absolute Deviation Plot

Usage

```
madPlot(hvt.scoring, ...)
```

Arguments

```
hvt.scoring List. A list of hvt.scoring obtained from the scoreHVT function.

... The ellipsis is passed to it as additional argument. (Used internally)
```

Details

This function plots percentage anomalies vs mean absolute deviation for test data. The plot helps in deciding an optimal MAD value for the use case.

Value

Mean Absolute Deviation Plot

mad_plot ggplot plot. A plot with percentage anomalies on y axis and mean absolute deviation values on xaxis.

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>

See Also

scoreHVT

Examples

```
data("EuStockMarkets")
dataset <- data.frame(date = as.numeric(time(EuStockMarkets)),</pre>
                      DAX = EuStockMarkets[, "DAX"],
                      SMI = EuStockMarkets[, "SMI"],
                      CAC = EuStockMarkets[, "CAC"],
                      FTSE = EuStockMarkets[, "FTSE"])
#adding this step especially for this function
rownames(EuStockMarkets) <- dataset$date</pre>
train <- EuStockMarkets[1:1302, ]</pre>
test <- EuStockMarkets[1303:1860, ]</pre>
hvt_summary <- list()</pre>
hvt_summary<- trainHVT(train,n_cells = 15, depth = 1, quant.err = 0.2,
                       distance_metric = "L1_Norm", error_metric = "mean",
                       projection.scale = 10, normalize = TRUE, seed = 123,
                       quant_method = "kmeans")
score_var <- scoreHVT(test, hvt_summary, child.level = 2, mad.threshold = 0.2)</pre>
madPlot(hvt.scoring=score_var)
```

plotAnimatedFlowmap

Generating flow maps and animations based on transition probabilities

Description

This is the main function for generating flow maps and animations based on transition probabilities including self states and excluding self states. Flow maps are a type of data visualization used to represent the transition probability of different states. Animations are the gifs used to represent the movement of data through the cells.

Usage

```
plotAnimatedFlowmap(
  hvt_model_output,
  transition_probability_df,
  df,
  animation = NULL,
  flow_map = NULL,
  fps_time = 1,
  fps_state = 1,
```

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```
time_duration = 2,
  state_duration = 2,
  cellid_column,
  time_column
)
```

Arguments

hvt_model_output

List. Output from a trainHVT function.

transition_probability_df

List. Output from getTransitionProbability function

df Data frame. The input dataframe should contain two columns, cell ID from

scoreHVT function and time stamp of that dataset.

animation Character. Type of animation ('state_based', 'time_based', 'All' or NULL) Character. Type of flow map ('self_state', 'without_self_state', 'All' or NULL)

fps_time Numeric. A numeric value for the frames per second of the time transition gif.

(Must be a numeric value and a factor of 100). Default value is 1.

fps_state Numeric. A numeric value for the frames per second of the state transition gif.

(Must be a numeric value and a factor of 100). Default value is 1.

 ${\tt time_duration} \quad Numeric. \ A \ numeric \ value \ for \ the \ duration \ of \ the \ time \ transition \ gif. \ Default$

value is 2.

state_duration Numeric. A numeric value for the duration of the state transition gif. Default

value is 2.

cellid_column Character. Name of the column containing cell IDs.

time_column Character. Name of the column containing time stamps

Value

A list of flow maps and animation gifs.

Author(s)

PonAnuReka Seenivasan <ponanureka.s@mu-sigma.com>, Vishwavani vishwavani@mu-sigma.com>

See Also

```
trainHVT
scoreHVT
getTransitionProbability
```

Examples

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plotHVT

Plot the hierarchical tessellations.

Description

This is the main plotting function to construct hierarchical voronoi tessellations in 1D,2D or Interactive surface plot.

Usage

```
plotHVT(
  hvt.results,
  line.width = 0.5,
  color.vec = "black",
  pch1 = 21,
  centroid.size = 1.5,
  title = NULL,
 maxDepth = NULL,
  child.level,
  hmap.cols,
  quant.error.hmap = NULL,
  cell_id = FALSE,
  n_{cells.hmap} = NULL,
  label.size = 0.5,
  sepration_width = 7,
  layer_opacity = c(0.5, 0.75, 0.99),
  dim_size = 1000,
  plot.type = "2Dhvt"
)
```

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Arguments

hvt.results	(2DProj/2Dhvt/2Dheatmap/surface_plot) List. A list containing the output of trainHVT function which has the details of the tessellations to be plotted.			
line.width	(2Dhvt/2Dheatmap) Numeric Vector. A vector indicating the line widths of the tessellation boundaries for each level.			
color.vec	(2Dhvt/2Dheatmap) Vector. A vector indicating the colors of the boundaries of the tessellations at each level.			
pch1	(2Dhvt/2Dheatmap) Numeric. Symbol of the centroids of the tessellations (parent levels). Default value is 21.			
centroid.size	(2Dhvt/2Dheatmap) Numeric. Size of centroids of first level tessellations.			
title	(2Dhvt) Character. Set a title for the plot. (default = NULL)			
maxDepth	(2Dhvt) Numeric. An integer indicating the number of levels. (default = NULL)			
child.level	(2Dheatmap/surface_plot) Numeric. Indicating the level for which the heat map is to be plotted.			
hmap.cols	(2Dheatmap/surface_plot) Numeric or Character. The column number or column name from the dataset indicating the variables for which the heat map is to be plotted.			
quant.error.hm	ар			
	(2Dheatmap) Numeric. A number indicating the quantization error threshold.			
cell_id	(2Dhvt) Logical. To indicate whether the plot should have Cell IDs or not for the first layer. (default = FALSE)			
n_cells.hmap	(2Dheatmap/surface_plot) Numeric. An integer indicating the number of cells/clusters per hierarchy (level)			
label.size	(2Dheatmap) Numeric. The size by which the tessellation labels should be scaled. (default = 0.5)			
sepration_width				
	(surface_plot) Numeric. An integer indicating the width between two levels			
layer_opacity	(surface_plot) Numeric. A vector indicating the opacity of each layer/ level			
dim_size	(surface_plot) Numeric. An integer indicating the dimension size used to create the matrix for the plot			
plot.type	Character. An option to indicate which type of plot should be generated. Accepted entries are '1D','2Dproj','2Dhvt','2Dheatmap'and 'surface_plot'. Default value is '2Dhvt'.			

Value

plot object containing the visualizations of reduced dimension(1D/2D) for the given dataset.

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>, Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also

trainHVT

Examples

plotModelDiagnostics Make the diagnostic plots for hierarchical voronoi tessellations

Description

This is the main function that generates diagnostic plots for hierarchical voronoi tessellations models and scoring.

Usage

```
plotModelDiagnostics(model_obj)
```

Arguments

model_obj

List. A list obtained from the trainHVT function or scoreHVT function

Value

For trainHVT, Minimum Intra-DataPoint Distance Plot, Minimum Intra-Centroid Distance Plot Mean Absolute Deviation Plot, Distribution of Number of Observations in Cells, for Training Data and Mean Absolute Deviation Plot for Validation Data are plotted. For scoreHVT Mean Absolute Deviation Plot for Training Data and Validation Data are plotted

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>

See Also

plotHVT

plotNovelCells 15

Examples

plotNovelCells

Plot the identified outlier cell(s) in the voronoi tessellation map.

Description

This is the main plotting function to construct hierarchical voronoi tessellations and highlight the outlier cells

Usage

```
plotNovelCells(
  plot.cells,
  hvt.map,
  line.width = c(0.6),
  color.vec = c("#141B41"),
  pch = 21,
  centroid.size = 0.5,
  title = NULL,
  maxDepth = 1
)
```

Arguments

plot.cells	Vector. A vector indicating the cells to be highlighted in the map
hvt.map	List. A list containing the output of trainHVT function which has the details of the tessellations to be plotted $$
line.width	Numeric Vector. A vector indicating the line widths of the tessellation boundaries for each level
color.vec	Vector. A vector indicating the colors of the boundaries of the tessellations at each level
pch	Numeric. Symbol of the centroids of the tessellations (parent levels) Default value is 21.
centroid.size	Numeric. Size of centroids of first level tessellations. Default value is 0.5
title	String. Set a title for the plot. (default = NULL)
maxDepth	Numeric. An integer indicating the number of levels. (default = NULL)

Value

Returns a ggplot object containing hierarchical voronoi tessellation plot highlighting the outlier cells

Author(s)

Shantanu Vaidya <shantanu.vaidya@mu-sigma.com>

See Also

```
trainHVT
plotHVT
```

Examples

plotQuantErrorHistogram

Make the quantization error plots for training and scoring.

Description

This is the function that produces histograms displaying the distribution of Quantization Error (QE) values for both train and test datasets, highlighting mean values with dashed lines for quick evaluation.

Usage

```
plotQuantErrorHistogram(hvt.results, hvt.scoring)
```

Arguments

```
hvt.results List. A list of hvt.results obtained from the trainHVT function.
hvt.scoring List. A list of hvt.scoring obtained from the scoreHVT function.
```

Value

Returns the ggplot object containing the quantization error distribution plots for the given HVT results of training and scoring

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Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>

See Also

plotHVT

Examples

plotStateTransition

Creating State Transition Plot

Description

This is the main function to create a state transition plot from a data frame. A state transition plot is a type of data visualization used to represent the changes or transitions in states over time for a given system. State refers to a particular condition or status of a cell at a specific point in time. Transition refers to the change of state for a cell from one condition to another over time.

Usage

```
plotStateTransition(
   df,
   sample_size = NULL,
   line_plot = NULL,
   cellid_column,
   time_column
)
```

Arguments

df	Data frame. The Input data frame should contain two columns. Cell ID from scoreHVT function and time stamp of that dataset.
sample_size	Numeric. An integer indicating the fraction of the data frame to visualize in the plot. Default value is 0.2
line_plot	Logical. A logical value indicating to create a line plot. Default value is NULL.
cellid_column	Character. Name of the column containing cell IDs.
time_column	Character. Name of the column containing time stamps.

Value

A plotly object representing the state transition plot for the given data frame.

Author(s)

PonAnuReka Seenivasan <ponanureka.s@mu-sigma.com>

See Also

trainHVT scoreHVT

Examples

```
reconcileTransitionProbability
```

Reconciliation of Transition Probability

Description

This is the main function for creating reconciliation plots and tables which helps in comparing the transition probabilities calculated manually and from markovchain function

Usage

```
reconcileTransitionProbability(
   df,
   hmap_type = NULL,
   cellid_column,
   time_column
)
```

Arguments

df Data frame. The input data frame should contain two columns, cell ID from scoreHVT function and timestamp of that dataset.

hmap_type Character. ('self_state', 'without_self_state', or 'All')

cellid_column Character. Name of the column containing cell IDs.

time_column Character. Name of the column containing timestamps

Value

A list of plotly heatmap objects and tables representing the transition probability heatmaps.

Author(s)

PonAnuReka Seenivasan <ponanureka.s@mu-sigma.com>, Vishwavani <vishwavani@mu-sigma.com>

See Also

trainHVT scoreHVT

Examples

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removeNovelty

Remove identified novelty cell(s)

Description

This function is used to remove the identified novelty cells.

Usage

```
removeNovelty(outlier_cells, hvt_results)
```

Arguments

outlier_cells Vector. A vector with the cell number of the identified novelty
hvt_results List. A list having the results of the compressed map i.e. output of trainHVT

function

Value

A list of two items

[[1]] Dataframe of novelty cell(s)

[[2]] Dataframe without the novelty cell(s) from the dataset used in model training

Author(s)

Shantanu Vaidya <shantanu.vaidya@mu-sigma.com>

See Also

```
trainHVT
scoreLayeredHVT
```

Examples

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scoreHVT

Score which cell each point in the test dataset belongs to.

Description

This function scores each data point in the test dataset based on a trained hierarchical Voronoi tessellations model.

Usage

```
scoreHVT(
   data,
   hvt.results.model,
   child.level = 1,
   mad.threshold = 0.2,
   line.width = c(0.6, 0.4, 0.2),
   color.vec = c("navyblue", "slateblue", "lavender"),
   normalize = TRUE,
   seed = 300,
   distance_metric = "L1_Norm",
   error_metric = "max",
   yVar = NULL
)
```

Arguments

data Data frame. A data frame containing the test dataset.

hvt.results.model

List. A list obtained from the trainHVT function

child.level Numeric. A number indicating the depth for which the heat map is to be plotted.

mad. threshold Numeric. A numeric value indicating the permissible Mean Absolute Deviation.

line.width Vector. A vector indicating the line widths of the tessellation boundaries for

each layer.

color.vec Vector. A vector indicating the colors of the tessellation boundaries at each

layer.

normalize Logical. A logical value indicating if the dataset should be normalized. When

set to TRUE, the data (testing dataset) is standardized by 'mean' and 'sd' of the training dataset referred from the trainHVT(). When set to FALSE, the data is

used as such without any changes.

seed Numeric. Random Seed to preserve the repeatability

distance_metric

Character. The distance metric can be L1_Norm(Manhattan) or L2_Norm(Eucledian). L1_Norm is selected by default. The distance metric is used to calculate the distance between an n dimensional point and centroid. The distance metric can be different from the one used during training.

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error_metric Character. The error metric can be mean or max. max is selected by default. max will return the max of m values and mean will take mean of m values where each value is a distance between a point and centroid of the cell.

yVar Character. A character or a vector representing the name of the dependent vari-

able(s)

Value

Dataframe containing scored data, plots and summary

Author(s)

Shubhra Prakash <shubhra.prakash@mu-sigma.com>, Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also

```
trainHVT
plotHVT
```

Examples

```
data("EuStockMarkets")
dataset <- data.frame(date = as.numeric(time(EuStockMarkets)),</pre>
                      DAX = EuStockMarkets[, "DAX"],
                      SMI = EuStockMarkets[, "SMI"],
                      CAC = EuStockMarkets[, "CAC"],
                      FTSE = EuStockMarkets[, "FTSE"])
rownames(EuStockMarkets) <- dataset$date</pre>
# Split in train and test
train <- EuStockMarkets[1:1302, ]</pre>
test <- EuStockMarkets[1303:1860, ]</pre>
#model training
hvt.results<- trainHVT(train,n_cells = 60, depth = 1, quant.err = 0.1,
                       distance_metric = "L1_Norm", error_metric = "max",
                       normalize = TRUE,quant_method = "kmeans")
scoring <- scoreHVT(test, hvt.results)</pre>
data_scored <- scoring$scoredPredictedData</pre>
```

scoreLayeredHVT

Score which cell and what layer each data point in the test dataset belongs to

Description

This function that scores the cell and corresponding layer for each data point in a test dataset using three hierarchical vector quantization (HVT) models (Map A, Map B, Map C) and returns a data frame containing the scored layer output. The function incorporates the scored results from each map and merges them to provide a comprehensive result.

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Usage

```
scoreLayeredHVT(
  data,
  hvt_mapA,
  hvt_mapB,
  hvt_mapC,
  mad.threshold = 0.2,
  normalize = TRUE,
  seed = 300,
  distance_metric = "L1_Norm",
  error_metric = "max",
  child.level = 1,
  yVar = NULL
)
```

Arguments

data	Data Frame.	A data f	frame containing	test dataset.	The data f	frame should have
------	-------------	----------	------------------	---------------	------------	-------------------

all the variable(features) used for training.

hvt_mapA A list of hvt.results.model obtained from trainHVT function while performing

'trainHVT()' on train data

hvt_mapB A list of hvt.results.model obtained from trainHVT function while performing

'trainHVT()' on data with novelty(s)

hvt_mapC A list of hvt.results.model obtained from trainHVT function while performing

'trainHVT()' on data without novelty(s)

mad. threshold Numeric. A number indicating the permissible Mean Absolute Deviation

normalize Logical. A logical value indicating if the dataset should be normalized. When

set to TRUE, the data (testing dataset) is standardized by 'mean' and 'sd' of the training dataset referred from the trainHVT(). When set to FALSE, the data is

used as such without any changes. (Default value is TRUE).

seed Numeric. Random Seed.

distance_metric

Character. The distance metric can be L1_Norm(Manhattan) or L2_Norm(Eucledian).

L1_Norm is selected by default. The distance metric is used to calculate the distance between an n dimensional point and centroid. The distance metric can be

different from the one used during training.

error_metric Character. The error metric can be mean or max. max is selected by default. max

will return the max of m values and mean will take mean of m values where each

value is a distance between a point and centroid of the cell.

child.level Numeric. A number indicating the level for which the heat map is to be plotted.

yVar Character. A character or a vector representing the name of the dependent vari-

able(s)

Value

Dataframe containing scored layer output

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Author(s)

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See Also

trainHVT plotHVT

Examples

```
data("EuStockMarkets")
dataset <- data.frame(date = as.numeric(time(EuStockMarkets)),</pre>
                      DAX = EuStockMarkets[, "DAX"],
                      SMI = EuStockMarkets[, "SMI"],
                      CAC = EuStockMarkets[, "CAC"],
                      FTSE = EuStockMarkets[, "FTSE"])
rownames(EuStockMarkets) <- dataset$date</pre>
train <- EuStockMarkets[1:1302, ]</pre>
test <- EuStockMarkets[1303:1860, ]</pre>
###MAP-A
hvt_mapA <- trainHVT(train, n_cells = 150, depth = 1, quant.err = 0.1,
                     distance_metric = "L1_Norm", error_metric = "max",
                     normalize = TRUE, quant_method = "kmeans")
identified_Novelty_cells <- c(127,55,83,61,44,35,27,77)</pre>
output_list <- removeNovelty(identified_Novelty_cells, hvt_mapA)</pre>
data_with_novelty <- output_list[[1]]</pre>
data_with_novelty <- data_with_novelty[, -c(1,2)]</pre>
hvt_mapB <- trainHVT(data_with_novelty,n_cells = 10, depth = 1, quant.err = 0.1,
                     distance_metric = "L1_Norm", error_metric = "max",
                     normalize = TRUE, quant_method = "kmeans")
data_without_novelty <- output_list[[2]]</pre>
### MAP-C
hvt_mapC <- trainHVT(data_without_novelty,n_cells = 135,</pre>
                     depth = 1, quant.err = 0.1, distance_metric = "L1_Norm",
                     error_metric = "max", quant_method = "kmeans",
                     normalize = TRUE)
##SCORE LAYERED
data_scored <- scoreLayeredHVT(test, hvt_mapA, hvt_mapB, hvt_mapC)</pre>
```

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trainHVT

Constructing Hierarchical Voronoi Tessellations

Description

This is the main function to construct hierarchical voronoi tessellations. This is done using hierarchical vector quantization(hvq). The data is represented in 2D coordinates and the tessellations are plotted using these coordinates as centroids. For subsequent levels, transformation is performed on the 2D coordinates to get all the points within its parent tile. Tessellations are plotted using these transformed points as centroids.

Usage

```
trainHVT(
  dataset,
  min_compression_perc = NA,
  n_{cells} = NA,
  depth = 1,
  quant.err = 0.2,
  projection.scale = 10,
  normalize = FALSE,
  seed = 279,
  distance_metric = c("L1_Norm", "L2_Norm"),
  error_metric = c("mean", "max"),
  quant_method = c("kmeans", "kmedoids"),
  scale_summary = NA,
  diagnose = FALSE,
  hvt_validation = FALSE,
  train_validation_split_ratio = 0.8
)
```

Arguments

dataset

Data frame. A data frame, with numeric columns (features) will be used for training the model.

min_compression_perc

Numeric. An integer, indicating the minimum compression percentage to be achieved for the dataset. It indicates the desired level of reduction in dataset size compared to its original size.

n_cells

Numeric. An integer, indicating the number of cells per hierarchy (level).

depth

Numeric. An integer, indicating the number of levels. A depth of 1 means no hierarchy (single level), while higher values indicate multiple levels (hierarchy).

quant.err

Numeric. A number indicating the quantization error threshold. A cell will only breakdown into further cells if the quantization error of the cell is above the defined quantization error threshold.

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projection.scale

Numeric. A number indicating the scale factor for the tessellations to visualize the sub-tessellations well enough. It helps in adjusting the visual representation of the hierarchy to make the sub-tessellations more visible.

normalize

Logical. A logical value indicating if the dataset should be normalized. When set to TRUE, scales the values of all features to have a mean of 0 and a standard deviation of 1 (Z-score).

seed

Numeric. A Random Numeric Seed to preserve the repeatability.

distance_metric

Character. The distance metric can be L1_Norm(Manhattan) or L2_Norm(Eucledian). L1_Norm is selected by default. The distance metric is used to calculate the distance between an n dimensional point and centroid.

error_metric

Character. The error metric can be mean or max. max is selected by default. max will return the max of m values and mean will take mean of m values where each value is a distance between a point and centroid of the cell.

quant_method

Character. The quantization method can be kmeans or kmedoids. Kmeans uses means (centroids) as cluster centers while Kmedoids uses actual data points (medoids) as cluster centers. kmeans is selected by default.

scale_summary

List. A list with user-defined mean and standard deviation values for all the features in the dataset. Pass the scale summary when normalize is set to FALSE.

diagnose

Logical. A logical value indicating whether user wants to perform diagnostics on the model. Default value is FALSE.

hvt_validation Logical. A logical value indicating whether user wants to holdout a validation set and find mean absolute deviation of the validation points from the centroid. Default value is FALSE.

train_validation_split_ratio

Numeric. A numeric value indicating train validation split ratio. This argument is only used when hvt_validation has been set to TRUE. Default value for the argument is 0.8

Value

[[4]]

A Nested list that contains the hierarchical tessellation information. This list has to be given as input argument to plot the tessellations.

[[1]]	A list containing information related to plotting tessellations. This information
	will include coordinates, boundaries, and other details necessary for visualizing
	the tessellations

[[2]] A list containing information related to Sammon's projection coordinates of the data points in the reduced-dimensional space.

[[3]] A list containing detailed information about the hierarchical vector quantized data along with a summary section containing no of points, Quantization Error and the centroids for each cell.

A list that contains all the diagnostics information of the model when diagnose is set to TRUE. Otherwise NA.

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[[5]]	A list that contains all the information required to generates a Mean Absolute Deviation (MAD) plot, if hvt_validation is set to TRUE. Otherwise NA
[[6]]	A list containing detailed information about the hierarchical vector quantized data along with a summary section containing no of points, Quantization Error and the centroids for each cell which is the output of 'hvq'
[[7]]	model info: A list that contains model-generated timestamp, input parameters passed to the model and the validation results

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See Also

plotHVT

Examples

VQ_codebookSplit

VQ_codebookSplit

Description

Vector Quantization by codebook split method

Usage

```
\label{eq:VQ_codebookSplit} VQ\_codebookSplit(dataset, quant.err = 0.5, epsilon = NULL)
```

Arguments

dataset	Matrix. A matrix of multivariate data. Each row corresponds to an observation, and each column corresponds to a variable. Missing values are not accepted.
quant.err	Numeric. The quantization error for the algorithm.
epsilon	Numeric. The value to offset the codebooks during the codebook split. Default is NULL, in which case the value is set to quant.err parameter.

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Details

Performs Vector Quantization by codebook split method. Initially, the entire dataset is considered to be one cluster where the codebook is the mean of the cluster. The quantization criteria is checked and the codebook is split such that the new codebooks are (codebook+epsilon) and (codebook-epsilon). The observations are reassigned to these new codebooks based on the nearest neighbour condition and the means recomputed for the new clusters. This is done iteratively until all the clusters meet the quantization criteria.

Value

clusters List. A list showing each ID assigned to a cluster. nodes.clust List. A list corresponding to nodes' details.

idnodes List. A list of ID and segments similar to nodes.clust with additional columns

for nodes ID.

error.quant List. A list of quantization error for all levels and nodes.

plt.clust List. A list of logical values indicating if the quantization error was met.

summary. Output table with summary.

Author(s)

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See Also

plotHVT

Examples

```
data("iris", package = "datasets")
iris <- iris[, 1:2]
vqOutput <- VQ_codebookSplit(iris, quant.err = 0.5)</pre>
```

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