Package ‘muHVT’

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getOptimalCentroids

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Description

Get Optimal Centroids

Usage

getOptimalCentroids(
  x,
  iter.max,
  algorithm,
  nclust,
  distance_metric,
  error_metric,
  quant.err
)

Arguments

x Data Frame. A data frame of multivariate data. Each row corresponds to an observation, and each column corresponds to a variable. Missing values are not accepted.
iter.max The max number of iterations the getOptimalCentroid function will run to get the optimal number of centroids
algorithm String. The type of algorithm used for quantization. Available algorithms are Hartigan and Wong, "Lloyd", "Forgy", "MacQueen". (default is "Hartigan-Wong")
nclust Numeric. Indicating the number of nodes per hierarchy.
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.err Numeric. The quantization error for the algorithm.
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.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**: Summary. Output table with summary.

Author(s)

Sangeet Moy Das <sangeet.das@mu-sigma.com>

Description

Hierarchical Vector Quantization

Usage

```r
hvq(
  x,
  nclust = 15,
  depth = 3,
  quant.err = 0.2,
  algorithm = c("Hartigan-Wong", "Lloyd", "Forgy", "MacQueen"),
  distance_metric = c("L1_Norm", "L2_Norm"),
  error_metric = c("mean", "max")
)
```
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.

- **nclust**: Numeric. Indicating the number of nodes per hierarchy.

- **depth**: Numeric. Indicating the hierarchy depth (or) the depth of the tree (1 = no hierarchy, 2 = 2 levels, etc.)

- **quant.err**: Numeric. The quantization error for the algorithm.

- **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

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.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**: Summary. Output table with summary.

Author(s)

Sangeet Moy Das <Sangeet.Das@mu-sigma.com>

See Also

hvtHmap
**Examples**

```r
data("USArrests", package="datasets")
hvqOutput = hvq(USArrests, nclust = 3, depth = 3, quant.err = 0.2,
                    distance_metric = 'L1_Norm', error_metric = 'mean')
```

---

**Description**

Main function to construct hierarchical voronoi tessellations.

**Usage**

```r
HVT(
    dataset,
    nclust = 15,
    depth = 3,
    quant.err = 0.2,
    projection.scale = 10,
    normalize = TRUE,
    distance_metric = c("L1_Norm", "L2_Norm"),
    error_metric = c("mean", "max")
)
```

**Arguments**

- **dataset**
  Data frame. A data frame with different columns is given as input.

- **nclust**
  Numeric. An integer indicating the number of clusters per hierarchy (level)

- **depth**
  Numeric. An integer indicating the number of levels. (1 = No hierarchy, 2 = 2 levels, etc ...)

- **quant.err**
  Numeric. A number indicating the quantization error threshold.

- **projection.scale**
  Numeric. A number indicating the scale factor for the tessellations so as to visualize the sub-tessellations well enough.

- **normalize**
  Logical. A logical value indicating if the columns in your dataset should be normalized. Default value is TRUE.

- **distance_metric**
  Character. The distance metric can be "Euclidean" or "Manhattan". Euclidean is selected by default.

- **error_metric**
  Character. The error metric can be "mean" or "max". mean is selected by default.
Details

This is the main function to construct hierarchical voronoi tessellations. The hvq function is called from this function. The output of the hvq function is hierarchical clustered data which will be the input for constructing tessellations. The data is then 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. The lines in the tessellations are chopped in places so that they do not protrude outside the parent polygon. This is done for all the subsequent levels.

Value

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

- [[1]] List. Information about the tesselation co-ordinates - level wise
- [[2]] List. Information about the polygon co-ordinates - level wise
- [[3]] List. Information about the hierarchical vector quantized data - level wise

Author(s)

Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also

plotHVT
hvtHmap

Examples

data(USArrests)
hvt.results <- list()
hvt.results <- HVT(USArrests, nclust = 15, depth = 1, quant.err = 0.2,
distance_metric = "L1_Norm", error_metric = "mean",
projection.scale = 10, normalize = TRUE)
plotHVT(hvt.results, line.width = c(0.8), color.vec = c('#141B41'),
maxDepth = 1)

hvt.results <- list()
hvt.results <- HVT(USArrests, nclust = 3, depth = 3, quant.err = 0.2,
distance_metric = "L1_Norm", error_metric = "mean",
projection.scale = 10, normalize = TRUE)
plotHVT(hvt.results, line.width = c(1.2,0.8,0.4), color.vec = c('#141B41','#0582CA','#8BA0B4'),
maxDepth = 3)
hvtHmap

Heat Map over Hierarchical Voronoi Tessellations

Description
Main function to construct heatmap overlay for hierarchical voronoi tessellations.

Usage

hvtHmap(
  hvt.results,  # List. A list of hvt.results obtained from the HVT function.
  dataset,     # Data frame. The input dataset.
  child.level, # Numeric. Indicating the level for which the heat map is to be plotted.
  hmap.cols,   # Numeric or Character. The column number of column name from the dataset
                # indicating the variables for which the heat map is to be plotted.
  color.vec = NULL, # Vector. A color vector such that length(color.vec) = (child.level - 1). (default = NULL)
  line.width = NULL, # Vector. A line width vector such that length(line.width) = (child.level - 1). (default = NULL)
  centroid.size = 3, # Numeric. Indicating the centroid size of the first level. (default = 3)
  pch = 21,        # Numeric. Indicating the centroid’s symbol type. (default = 21)
  palette.color = 6, # Numeric. Indicating the heat map color palette. 1 - rainbow, 2 - heat.colors, 3 - terrain.colors, 4 - topo.colors, 5 - cm.colors, 6 - seas color. (default = 6)
  previous_level_heatmap = TRUE,  # previous_level_heatmap = TRUE,
  show.points = FALSE,             # show.points = FALSE,
  asp = 1,                        # asp = 1,
  ask = TRUE,                     # ask = TRUE,
  tess.label = NULL,              # tess.label = NULL,
  quant.error.hmap = NULL,        # quant.error.hmap = NULL,
  nclust.hmap = NULL,             # nclust.hmap = NULL,
  label.size = 0.5,              # label.size = 0.5,
  ...)

Arguments

- **hvt.results**: List. A list of hvt.results obtained from the HVT function.
- **dataset**: Data frame. The input data set.
- **child.level**: Numeric. Indicating the level for which the heat map is to be plotted.
- **hmap.cols**: Numeric or Character. The column number of column name from the dataset indicating the variables for which the heat map is to be plotted.
- **color.vec**: Vector. A color vector such that length(color.vec) = (child.level - 1). (default = NULL)
- **line.width**: Vector. A line width vector such that length(line.width) = (child.level - 1). (default = NULL)
- **centroid.size**: Numeric. Indicating the centroid size of the first level. (default = 3)
- **pch**: Numeric. Indicating the centroid’s symbol type. (default = 21)
- **palette.color**: Numeric. Indicating the heat map color palette. 1 - rainbow, 2 - heat.colors, 3 - terrain.colors, 4 - topo.colors, 5 - cm.colors, 6 - seas color. (default = 6)
previous_level_heatmap

Logical. If TRUE, the heatmap of previous level will be overlayed on the heatmap of selected level. If FALSE, the heatmap of only selected level will be plotted.

show.points

Logical. Indicating if the centroids should be plotted on the tessellations. (default = FALSE)

asp

Numeric. Indicating the aspect ratio type. For flexible aspect ratio set, asp = NA. (default = 1)

ask

Logical. If TRUE (and the R session is interactive) the user is asked for input, before a new figure is drawn. (default = TRUE)

tess.label

Vector. A vector for labelling the tessellations. (default = NULL)

quant.error.hmap

Numeric. A number indicating the quantization error threshold.

nclust.hmap

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

label.size

Numeric. The size by which the tessellation labels should be scaled. (default = 0.5)

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

Details

The output of the HVT function has all the required information about the HVT. Now a heat map is overlayed over this HVT. The user defines the level and also those variables of the data for which the heat map is to be plotted. Now for each variable a separate heat map is plotted. The plot area is divided into 2 screens where the first screen is relatively large and will have the heat map. The second screen is small and contains the gradient scale. To plot the heat map, the data is first normalized. The gradient scale is divided into 'n' regions (500 is the set default). Using the normalized data, the different regions into which the data items fall are found. Each data item is now having a region on the gradient scale. This color is filled in the tile corresponding to the data item. This procedure is done for all the tiles for that level to get the complete heat map. Once the heat map is ready, the higher level tessellations are plotted to represent the hierarchies. The size of the centroids, the thickness of the lines and the color of the tessellation lines can be given as input by the user. Appropriate values for these parameters should be given to identify the hierarchies properly. In the second screen the gradient scale is plotted. The heat maps and hierarchical tessellations are obtained for all the desired variables.

Author(s)

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See Also

plotHVT

Examples

data(USArrests)
hvt.results <- list()
plotHVT <- HVT(USArrests, nclust = 6, depth = 1, quant.err = 0.2,
                distance_metric = "L1_Norm", error_metric = "mean",
                projection.scale = 10, normalize = TRUE)

hvtHmap(hvt.results, USArrests, child.level = 1,hmap.cols = 'Murder',
        line.width = c(0.2), color.vec = c('#141B41'),palette.color = 6,
        quant.error.hmap = 0.2,nclust.hmap = 6)

hvt.results <- list()

hvt.results <- HVT(USArrests, nclust = 3, depth = 3, quant.err = 0.2,
                    distance_metric = "L1_Norm", error_metric = "mean",
                    projection.scale = 10, normalize = TRUE)

hvtHmap(hvt.results, USArrests, child.level = 3,hmap.cols = 'Quant.Error',
        line.width = c(1.2,0.8,0.4),color.vec = c('#141B41', '#0582CA', '#8BA0B4'),
        palette.color = 6,quant.error.hmap = 0.2,nclust.hmap = 3)

---

**plotHVT**

*Plot the hierarchical tessellations.*

**Description**

Main plotting function to construct hierarchical voronoi tessellations.

**Usage**

```r
plotHVT(
  hvt.results,
  line.width,
  color.vec,
  pch1 = 21,
  centroid.size = 3,
  title = NULL,
  maxDepth = NULL
)
```

**Arguments**

- **hvt.results** List. A list containing the output of HVT 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.
- **pch1** Numeric. Symbol type of the centroids of the tessellations (parent levels). Refer `points`. (default = 21)
- **centroid.size** Numeric. Size of centroids of first level tessellations. (default = 3)
- **title** String. Set a title for the plot. (default = NULL)
- **maxDepth** Numeric. An integer indicating the number of levels. (default = NULL)
**predictHVT**

*Predict which cell and what level each point in the test dataset belongs to*

**Description**

Main function to predict cell path of new datapoints

**Usage**

```r
predictHVT(
  data, 
  hvt.results, 
  hmap.cols = NULL, 
  child.level = 1, 
  quant.error.hmap = NULL, 
  nclust.hmap = NULL, 
  line.width = NULL, 
  color.vec = NULL, 
  ...
)
```

**Arguments**

- **data** List. A dataframe containing test dataset. The dataframe should have atleast one variable used while training. The variables from this dataset can also be used to overlay as heatmap
predictHVT

hvt.results  A list of hvt.results obtained from HVT function while performing hierarchical vector quantization on train data
hmap.cols    - The column number of column name from the dataset indicating the variables for which the heat map is to be plotted.(Default = # NULL). A heatmap won’t be plotted if NULL is passed
child.level  A number indicating the level for which the heat map is to be plotted.(Only used if hmap.cols is not NULL)
quant.error.hmap  Numeric. A number indicating the quantization error threshold.
nclust.hmap   Numeric. An integer indicating the number of clusters per hierarchy;line.width  Vector. A line width vector such that length(line.width) = (child.level - 1). (default = NULL)
color.vec     Vector. A color vector such that length(color.vec) = (child.level - 1). (default = NULL)
...  color.vec and line.width can be passed from here

Author(s)
Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also
HVT
hvtHmap

Examples

data(USArrests)
#Split in train and test

train <- USArrests[1:40,]
test <- USArrests[41:50,]

hvt.results <- list()
hvt.results <- HVT(train, nclust = 3, depth = 2, quant.err = 0.2,
  distance_metric = "L1_Norm", error_metric = "mean",
  projection.scale = 10, normalize = TRUE)

predictions <- predictHVT(test,hvt.results,hmap.cols = "Quant.Error", child.level=2,
  quant.error.hmap = 0.2,nclust.hmap = 3,line.width = c(1.2,0.8,0.4),
  color.vec = c('#141B41','#0582CA','#8BA0B4'))
print(predictions$predictions)
Description

This is a wrapper for the sammon function of the MASS package for non-metric multidimensional scaling.

Usage

```r
sammonsProjection(
  d,
  y = stats::cmdscale(d, k),
  k = 2,
  niter = 100,
  trace = TRUE,
  magic = 0.2,
  tol = 1e-04
)
```

Arguments

- `d`: distance structure of the form returned by dist, or a full, symmetric matrix. Data are assumed to be dissimilarities or relative distances, but must be positive except for self-distance. This can contain missing values.
- `y`: An initial configuration. If none is supplied, cmdscale is used to provide the classical solution. (If there are missing values in `d`, an initial configuration must be provided.) This must not have duplicates.
- `k`: The dimension of the configuration.
- `niter`: The maximum number of iterations.
- `trace`: Logical for tracing optimization. Default TRUE.
- `magic`: initial value of the step size constant in diagonal Newton method.
- `tol`: Tolerance for stopping, in units of stress.

Details

This chooses a two-dimensional configuration to minimize the stress, the sum of squared differences between the input distances and those of the configuration, weighted by the distances, the whole sum being divided by the sum of input distances to make the stress scale-free.

An iterative algorithm is used, which will usually converge in around 50 iterations. As this is necessarily an O(n^2) calculation, it is slow for large datasets. Further, since the configuration is only determined up to rotations and reflections (by convention the centroid is at the origin), the result can vary considerably from machine to machine. In this release the algorithm has been modified by adding a step-length search (magic) to ensure that it always goes downhill.
Value

points A two-column vector of the fitted configuration.
stress The final stress achieved.

Examples

```r
require(MASS)
swiss.x <- as.matrix(swiss[, -1])
swiss.sam <- sammonsProjection(dist(swiss.x))
```

Description

Vector Quantization by codebook split method

Usage

```r
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.

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.
VQ_codebookSplit

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.

Author(s)

Sangeet Moy Das <sangeet.das@mu-sigma.com>

See Also

hvtHmap

Examples

data("iris", package="datasets")
iris <- iris[,1:2]

vqOutput = VQ_codebookSplit(iris, quant.err = 0.5)
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