# Package 'UMAPVizR'

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Type Package

**Title** Visualization and statistical analysis of high dimensional cytometry data using manifold representations

#### **Description**

Cytometry data are now classically analyzed using non linear dimensionality reduction approaches, but it is still challenging to easily handle the whole pipeline of computational analyses. UMAPVizR allows the statistical analysis and visualization of high dimensional cytometry data using manifold algorithms and clustering methods. Especially, several key analysis steps are available to perform data importation, manifold generation, cell cluster identification, statistical analyses, cluster visualization, and quality controls of generated results. UMAPVizR can import cell events from FCS or txt file formats using different transformation, down-sampling, and normalization approaches.

Manifold representations can be generated using the UMAP, tSNE or LargeVis algorithms to project cell events into a lower dimensionality space. The identification of cell clusters can be done using multiple clustering algorithms, depending on user's assumptions. The characteristics of cell clusters can be visualized using scatter plot, categorical heatmap of marker expressions, or using parallel coordinates representations. Cell clusters having abundances differently expressed between biological conditions can be identified using several statistical tests.

Statistical results can be visualized using volcano plots or heatmaps.

Unsupervised and supersized analysis approaches can be conducted by users to appreciate the homogeneity/heterogeneity of biological conditions, and to identify cell population biomarkers in a multivariate manner.

Additionally, UMAPVizR provides a workflow for asserting the quality of identified cell clusters using statistical approaches.

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#### Imports checkmate,

cluster,

concaveman,

cowplot,

dbscan,

dendextend,

diptest,

FactoMineR,

flowCore,

FNN,

2 R topics documented:

fpc, ggdendro, Gmedian, ggnewscale, ggplot2, ggpubr, ggrepel, ggridges, gridExtra, gtools, kohonen, MASS, methods, plyr, reshape, reshape2, rstatix, Rtsne, scales, stats, stringr, uwot Suggests knitr License GPL-3 | file LICENSE VignetteBuilder knitr

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# R topics documented:

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 $assign {\tt Metadata}$ 

Assigns meta-information about biological samples

## **Description**

This function aims to attach meta-information to each biological sample.

Especially, the biological individual, the biological condition and the time point of each sample can be specified for subsequent analyses.

## Usage

assignMetadata(UMAPdata, metadata)

## Arguments

UMAPdata a UMAPdata object

metadata a data.frame containing contextual information about the biological samples.

This data.frame must have 3 columns specifying for each sample the associated individual (column named 'individual'), the biological condition (column

named 'condition') and the time point (column named 'timepoint')

#### Value

a S4 object of class 'UMAPdata'

4 export

computeStatistics

Computes differential analysis statistics for cell clusters

## Description

This function aims to compute the statistics of Differentially Abundant Clusters 'DAC'.

DAC correspond to cell clusters having abundances statistically different between two biological conditions. The statistical test used for the comparisons can be defined by users. For each cluster, the p-value, log2 fold-change and effect size relative to the reference condition are computed. Statistical comparison can be performed in a paired and unpaired manner.

#### Usage

```
computeStatistics(
  UMAPdata,
  condition,
  ref.condition,
  test.statistics = c("wilcox.test", "t.test"),
  paired = FALSE
)
```

## **Arguments**

UMAPdata a UMAPdata object

condition a character value providing the name of the condition to be compared

ref.condition a character value providing the name of reference condition

test.statistics

a character value providing the type of statistical test to use. Possible values are:

'wilcoxon' or 't-test'

paired a boolean value indicating if a paired or unpaired comparison should be applied

#### Value

a S4 object of class 'UMAPdata'

export

Exports cell expression profiles to TSV or FCS files

## **Description**

Exports cell expresion profiles from a UMAPdata object to a tab-separated or FCS files.

Cell expression profiles can be exported for a set of given samples and for a set of given cell clusters.

## Usage

```
export(UMAPdata, filename, clusters = NULL, samples = NULL)
## S4 method for signature 'UMAPdata'
export(UMAPdata, filename, clusters = NULL, samples = NULL)
```

generateManifold 5

#### **Arguments**

UMAPdata a UMAPdata object

filename a character value providing the name of the output file

clusters a character vector containing the identifiers of the cell clusters to export. By

default, all clusters are extracted

samples a character vector containing the names of biological samples to export. By

default, all samples are extracted

#### Value

none

generateManifold

Generates a manifold of cell events

## **Description**

This function aims to generate a manifold representation for cell events stored in a UMAPdata object.

This function allows the use of several non-linear dimension reduction techniques such as UMAP, t-SNE or LargeVis. The whole set of cell markers or specific cell markers can be used during the dimensionality reduction process.

## Usage

```
generateManifold(
   UMAPdata,
   type = c("UMAP", "tSNE", "lvish"),
   markers = NULL,
   seed = 42,
   verbose = TRUE,
   ...
)
```

#### **Arguments**

UMAPdata a UMAPdata object

type a character value specifying the type of manifold to compute. Possible values

are: 'UMAP' for Uniform Manifold Approximation and Projection, 'tSNE' for

t Stochastic Neighbor Embedding, and 'lvish' for LargeVis

markers a character vector providing the cell markers to use for the manifold generation seed a numeric value providing the random seed to use during stochastic operations

verbose a boolean value indicating if computational details must be displayed on the

console

... Other arguments passed on to methods

#### Value

a S4 object of class 'UMAPdata'

6 identifyClusters

identifyClusters

Identify cell cluster of having similar marker expressions

#### **Description**

This function aims to identify cell clusters, which are groups of cells having similar expressions for selected markers, using different unsupervised clustering methods.

Several clustering method are available such as kmeans, kmedian, clara, DBSCAN, HDBSCAN and SOM. The cell clustering can be performed on the manifold representation or based on marker expression.

## Usage

```
identifyClusters(
  UMAPdata,
  space = c("manifold", "markers"),
  method = c("kmeans", "kmedian", "clara", "DBSCAN", "SOM"),
  concavity = 2,
  length.threshold = 0,
  seed = 42,
  ...
)
```

## **Arguments**

UMAPdata a UMAPdata object

space a character value containing the space of clustering method to use. Possible

values are: 'manifold' or 'markers'

method a character value containing the name of the clustering method to use. Possible

values are: 'kmeans', 'kmedian', 'clara', 'DBSCAN' and 'SOM'

concavity a numeric value providing a relative measure of concavity for the computation

of the concave hulls (please refer to the function 'concaveman' of the 'concave-

man' package)

length.threshold

a numeric value providing a threshold of the segment length for the computation of the concave hulls (please refer to the function 'concaveman' of the concave-

man package)

seed a numeric value providing the random seed to use during stochastic operations

... Other arguments passed on to methods

#### **Details**

For each identify cell cluster, the boundaries of cells belonging to this cluster are delineated using a concave hull

```
a S4 object of class 'UMAPdata'
```

import 7

import

Imports of cell expression profiles from TSV or FCS files

## Description

This function aims to import acquired cell events into a UMAPdata object.

Input files can be tab-separated or FCS files. Different transformations can be applied such as logicle, arcsinh or logarithmic. Importantly, a downsampling of cell events can be performed using uniformly-based or density-based random selections. Cell marker having technical or biological biaises can be excluded during the import.

## Usage

```
import(
  files,
  filetype = "fcs",
  transform = c("logicle", "arcsinh", "logarithmic", "none"),
  downsampling = NULL,
  d.method = c("uniform", "density"),
  exclude.markers = NULL,
  seed = 42
)
```

# Arguments

files	a character vector specifying the path of the tab-separated or FCS files to load							
filetype	a character vector specifying the format of the loaded files. By default, FCS is used $$							
transform	a character value containing the type of the transformation to apply. Possible values are: 'logicle', 'arcsinh', 'logarithmic' or 'none'							
downsampling	a numeric value providing the number of cells to downsample for each sample							
d.method	a character value containing the type of the downsampling to apply. Possible values are: 'uniform' or 'density'							
exclude.markers	exclude.markers							
	a character vector providing the marker names to be excluded during the import							
seed	a numeric value providing the random seed to use during stochastic operations							

```
a S4 object of class 'UMAPdata'
```

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performUpsampling

Performs the upsampling of downsampled events

## **Description**

This function aims to perform upsample downsampled events based on an existing UMAPdata object and existing cell events stored in tab-separated or FCS files.

Importantly, the identification of cell clusters must have been performed prior to this operation.

## Usage

```
performUpsampling(UMAPdata, files)
```

#### **Arguments**

UMAPdata a U

a UMAPdata object

files

a character vector providing the path of the tab-separated or FCS files

#### Value

a S4 object of class 'UMAPdata'

plot

Plots graphics for all UMAPdata objects

## Description

Generates a graphical representation for a UMAPdata object. The displayed representation depends on the current analysis status of the UMAPdata object.

- If the manifold has not been calculated, then the number of cells per sample will be displayed.
- If the manifold has been calculated but not the clustering, then the manifold representation will be displayed.
- If the manifold and clustering have been calculated, then a heatmap of marker expressions will be displayed.

## Usage

```
## S4 method for signature 'UMAPdata,ANY'
plot(x)
```

# **Arguments**

Х

a UMAPdata object

```
a ggplot2 object
```

plotBoxplot 9

plotBoxplot	Plots cell cluster abundances using a boxplot representation

# Description

This function aims to visualize and compare the cell cluster abundances for each biological condition using boxplot and jitter representations.

The abundance of a specific cell cluster or a set of cell clusters can be displayed. The representation can be resticted to a specific set of samples. Moreover boxplot can be constructed based on sample meta information. Statistic can be computed for all comparisons.

## Usage

```
plotBoxplot(
   UMAPdata,
   clusters,
   samples = NULL,
   observation = c("individual", "condition", "timepoint"),
   test.statistics = c("wilcox.test", "t.test"),
   paired = FALSE,
   hide.ns = TRUE
)
```

## **Arguments**

**UMAPdata** a UMAPdata object a character vector containing the identifiers of the clusters to use clusters a character vector containing the names of biological samples to use. By default, samples all samples are used observation a character value containing the parameters to use test.statistics a character value providing the type of statistical test to use. Possible values are: 'wilcox.test' or 't.test' paired a boolean value indicating if a paired or unpaired comparison should be applied hide.ns a boolean value indicating xxx

```
a ggplot2 object
```

10 plotCellCounts

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Plots the numbers of cells for each sample

## **Description**

This function aims to visualize the number of cells associated to each sample.

This representation displays the samples in the X-axis and the number of associated cells in the Y-axis. Several statistics can be computed and shown.

# Usage

```
plotCellCounts(
   UMAPdata,
   stats = c("min", "median", "mean", "q75", "max"),
   samples = NULL,
   sort = TRUE
)
```

## **Arguments**

umapdata a Umapdata object

stats a character vector providing the statistics to display. Possible values are: 'min', 'median', 'mean', 'q75', 'max'

samples a character vector containing the names of biological samples to use. By default, all samples are used

sort a boolean value indicating if clusters must be sorted by the number associated sample

#### **Details**

The following statistic can be computed:

- -'min' corresponds to the lowest number of cells within a data set
- -'median' corresponds to the number of cells separates the lower half from the upper half within data set
- -'mean' corresponds to the number of cells quantity shared within data set
- -'q75' corresponds to the number of cells separates the quantiles 75
- -'max' corresponds to the largest number of cells within a data set

```
a ggplot2 object
```

plotClustersCounts 11

р	lotClustersCounts	Plots the	numbers	of cells	for each	clusters
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## **Description**

This function aims to visualize the number of cells associated to each clusters.

This representation displays the clusters in the X-axis and the total number of associated cells in the Y-axis.

## Usage

```
plotClustersCounts(UMAPdata, clusters = NULL, sort = TRUE)
```

## **Arguments**

UMAPdata a UMAPdata object

clusters a character vector containing the identifiers of the clusters to use. By default, all

clusters are used

sort a boolean value indicating if clusters must be sorted by the number associated

cluster

#### Value

a ggplot2 object

plotCombineHM Plots a combined expression and statistic heatmaps

# Description

This function aims to combine the expression and statistic heatmaps.

#### Usage

```
plotCombineHM(HM1, HM2)
```

# Arguments

HM1 a ggplot object containing the expression heatmap
HM2 a ggplot object containing the statistic heatmap

```
a ggplot2 object
```

12 plotDistogram

plotCoordinates

Plots parallel coordinates

## **Description**

This function aims to visualize xxx

## Usage

```
plotCoordinates(UMAPdata)
```

## Arguments

UMAPdata

a UMAPdata object

#### Value

XX

plotDistogram

Plots of a distogram of marker co-expression

# Description

This function aims to visualize the pairwise co-expression between all markers using a distogram representation. Each tile corresponds to the co-expression between two markers and is gradient-colored based on the Pearson correlation.

## Usage

```
plotDistogram(UMAPdata, clusters = NULL)
```

## **Arguments**

UMAPdata a UMAPdata object

clusters a character vector containing the identifier of the cluster to use

#### **Details**

The Pearson correlation is computed based on the marker expressions.

```
a ggplot2 object
```

plotHmAbundances 13

plotHmAbundances	Plots an heatmap of cell cluster abundances	
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#### **Description**

This function aims to visualize the abundances of cell clusters using an heatmap representation.

In such heatmap each column corresponds a cell cluster and he row corresponds the different samples. The heatmap can be restricted to specific cell clusters and samples. The levels of abundance of each sample in each cluster is represented using a color gradient scale. Abundance values can be centered and reduced.

## Usage

```
plotHmAbundances(
   UMAPdata,
   clusters = NULL,
   samples = NULL,
   saturation = 2.5,
   rescale = FALSE
)
```

#### **Arguments**

UMAPdata a UMAPdata object

clusters a character vector containing the identifiers of the clusters to use. By default, all clusters are used

samples a character vector containing the names of biological samples to use. By default, all samples are used

saturation a numeric value providing the saturation threshold of cell cluster abundances

a boolean specifying if cell cluster abundances must be centered and reduced

#### Value

a ggplot2 object

rescale

plotHmExpressions Plots an heatmap of cell marker expressions

#### **Description**

This function aims to visualize the cell marker expressions for selected markers and clusters.

The mean of median marker expressions is computed for each cluster, and marker expressions displayed using a categorical heatmap (5 categories are defined by default). The range expression of each cell marker is discretized into several categories between bounds of marker expressions. To hierarchical clustering, shown using dendrogramm, can be computed on both marker and cluster levels.

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#### **Usage**

```
plotHmExpressions(
   UMAPdata,
   markers = NULL,
   clusters = NULL,
   method.hclust = c("ward.D", "ward.D2", "single", "complete", "average", "mcquitty",
        "median", "centroid"),
   nb.cat = 5,
   seed = 42
)
```

#### **Arguments**

UMAPdata a UMAPdata object

markers a character vector providing the marker names to use. By default, all markers

are used

clusters a character vector containing the identifiers of the clusters to use. By default, all

clusters are used

method.hclust a character value providing the agglomeration method to be use. Possible values

are: 'ward.D', 'ward.D2', 'single', 'complete', 'average', 'mcquitty', 'median'

or 'centroid' (please refer to the function 'hclust' of the 'stats' package)

nb.cat a numeric specifying the number of categories to use

seed a numeric value providing the random seed to use during stochastic operations

# Value

a ggplot2 object

plotHmStatistics

Plots an heatmap of a statistical analysis results

#### **Description**

This function aims to visualize the results of differential cell clusters analysis.

This representation displays statistical information for each cell cluster for a given comparison of samples. Different statistics can be visualized, such as the p-value, the log2(fold-change), and effect size.

# Usage

```
plotHmStatistics(
   UMAPdata,
   clusters = NULL,
   statistics = c("pvalue", "lfc", "effsize"),
   saturation = 3
)
```

plotLDA 15

#### **Arguments**

UMAPdata a UMAPdata object

clusters a character vector containing the identifiers of the clusters to use. By default, all

clusters are used

statistics a character value providing the name of the statistic to display. Possible values

are: 'pvalue' for p-value, 'lfc' for log2 fold change or 'eff' for effect size

saturation a numeric value providing the saturation value for statistics to display

#### Value

a ggplot2 object

plotLDA

Plots a LDA representation based cell cluster abundances

#### **Description**

This function aims to represent a Linear Discriminant Analysis representation based on cell cluster abundances.

#### Usage

plotLDA(UMAPdata)

## **Arguments**

UMAPdata a UMAPdata object

#### Value

XX

plotManifold Plots a representati

Plots a representation of a computed manifold

## **Description**

This function aims to visualize a computed manifold representation for given analysis.

This representation can be used on a UMAPdata object for which a manifold analysis has been performed.

If a cell clustering has been performed, then the clusters are delineated using concave hulls. Additionnally, the manifold can be colored based on the local cell density or marker expressions. It is possible to centred ans reduce the values of expressions.

16 plotMDS

#### Usage

```
plotManifold(
   UMAPdata,
   markers = "density",
   samples = NULL,
   scale = FALSE,
   quant.low = 0.05,
   quant.high = 0.95
)
```

#### **Arguments**

umarkers a character value providing the name of the marker to use for the colouring. By default, cells are colored based on their local density

samples a character vector containing the names of biological samples to use

scale a boolean value specifing if expression calue must be rescaled

quant.low a numeric value providing the number of first quantile

quant.high a numeric value providing the number of last quantile

#### Value

a ggplot2 object

plotMDS

Plots a MDS representation based on cell cluster abundances

## **Description**

This function aims to visualize the similarities between samples or clusters based on their abundances, using a Multidimensional Scaling representation. Each dot represents a sample or a cluster and the distances between the dots are proportional to the Euclidean distance between these objects. The representation can be restricted to specific cell clusters and samples. In addition, it is possible to choose the levels displayed, clusters or samples.

# Usage

```
plotMDS(
   UMAPdata,
   levels = c("clusters", "samples"),
   condition.samples = c("condition", "timepoint"),
   clusters = NULL,
   samples = NULL
)
```

plotPCA 17

#### **Arguments**

UMAPdata a UMAPdata object

levels a character value containing the variable to be displayed. Possible values are:

'clusters' or 'samples'

condition.samples

a character vector containing the variable to be studied for the samples. Possible

values are: 'condition' or 'timepoint"

clusters a character vector containing the identifiers of the clusters to use. By default, all

clusters are used

samples a character vector containing the names of biological samples to use. By default,

all samples are used

#### Value

a ggplot2 object

plotPCA

Plots a PCA representation based cell cluster abundances

#### **Description**

This function aims to represent a Principal Component Analysis representation based on cell cluster abundances. In such representation, clusters or samples are positioned based on computed principal components. The representation can be displayed based on specific principal components. The representation can be restricted to specific cell clusters and samples. In addition, it is possible to choose the levels displayed, clusters or samples.

#### Usage

```
plotPCA(
   UMAPdata,
   levels = c("both", "clusters", "samples"),
   clusters = NULL,
   samples = NULL,
   components = c(1, 2),
   condition.samples = c("condition", "timepoint"),
   cor.radius.th = 0.6,
   cluster.label.size = 3,
   sample.label.size = 3,
   cluster.dot.size = 2,
   sample.dot.size = 3
)
```

#### **Arguments**

UMAPdata a UMAPdata object

levels a character value containing the variable to be displayed. Possible values are:

'both', 'clusters' or 'samples'

18 plotPhenoClusters

clusters a character vector containing the identifier of the cluster to use. By default, all clusters are used
samples a character vector containing the names of biological samples to use. By default,

all samples are used

components a numeric vector providing the components to display

condition.samples

a character vector containing the variable to be studied for the samples. Possible

values are: 'condition' or 'timepoint"

cor.radius.th a numeric value specifing xxx

cluster.label.size

a numeric value specifing the label size for clusters

sample.label.size

a numeric value specifing the label size for the samples

cluster.dot.size

a numeric value specifing the dot size for the clusters

sample.dot.size

a numeric value specifing the dot size for the samples

#### Value

a ggplot2 object

plotPhenoClusters

Plots of phenotype of identified cell clusters

## **Description**

This function aims to visualize xxx

## Usage

```
plotPhenoClusters(
   UMAPdata,
   cluster,
   quant.low = 0.05,
   quant.high = 0.95,
   dip.th = 0.01
)
```

## **Arguments**

UMAPdata a UMAPdata object

cluster a character vector containing the identifier of the cluster to use

quant.low a numeric value providing the number of first quantile quant.high a numeric value providing the number of last quantile

dip.th a numeric value specifing xxx

#### Value

a ggplot2 object

plotScatter 19

plotScatter	Plots of a scatter plot of marker co-expression	

## **Description**

This function aims to visualize co-expression between two markers using a scatter representation

#### Usage

```
plotScatter(UMAPdata, marker1, marker2, samples = NULL, clusters = NULL)
```

## **Arguments**

UMAPdata	a UMAPdata object
marker1	a character value specifying the first marker to be visualised
marker2	a character value specifying the second marker to be visualised
samples	a character vector containing the names of biological samples to use. By default, all samples are used
clusters	a character vector containing the identifiers of the clusters to use. By default, all clusters are used

#### Value

a ggplot2 object

plotVolcanoPlot	Plots of a volcano plot of statistical analysis	

#### **Description**

This function aims to visualize the results of a differentially abundant a analysis using a Volcano plot In such representation, each in dot corresponds to a cell cluster and dots are positioned in two dimensional space where the X-axis represents the log2(fold-change) and the Y-axis represents the -log10 of the p-value. Un horizontal line is displayed accordingly to the p-value threshold and to vertical lines are displayed accordingly to the fold-change threshold.

## Usage

```
plotVolcanoPlot(UMAPdata, comparison, th.pv = 1.3, th.fc = 1.5)
```

## **Arguments**

```
UMAPdata a UMAPdata object comparison a character value containing
```

comparison a character value containing the comparison to study
th.pv a numeric value containing the p-value threshold to use
th.fc a numeric value containing the fold-change threshold to use

```
a ggplot2 object
```

20 QCMarkerNames

print

Prints information for a given UMAPdata object

## Description

Prints a preview for a UMAPdata object.

## Usage

```
## S4 method for signature 'UMAPdata'
print(x)
```

# Arguments

Χ

a UMAPdata object

#### Value

none

QCMarkerNames

Verifies the consistency of the marker names within cell event files

# Description

This function aims to check the consistency of marker names across multiple tab-separated or FCS files.

Additionally, the number of cells associated to each sample is displayed.

# Usage

```
QCMarkerNames(files)
```

## **Arguments**

files

a character vector specifying the path of the tab-separated or FCS files to check

#### Value

a data.frame containing the marker names and the associated number of cells for each sample (rownames = samples and colnames = markers)

QCMarkerRanges 21

QCMarkerRanges Verifies the consistency of marker expressions integrity event files	within cell
---	-------------

## **Description**

This function aims to check the consistency of marker expressions ranges across multiple tabseparated or FCS files.

The marker expressions ranges are calculated based on the user-defined quantiles.

## Usage

```
QCMarkerRanges(files, probs = c(0.05, 0.95))
```

## **Arguments**

files a character vector specifying the path of the FCS files to verified

probs a numerical vector providing the quantiles used to define marker expressions

ranges

## Value

a list containing two data.frame for the lower and upper marker expression ranges (rownames = samples and colnames = markers)

QCSmallClusters Computes the percentage of cell clusters with low number of cells

## **Description**

This function aims to compute and show cell clusters having a number of associated cells lower than a specific threshold

#### Usage

```
QCSmallClusters(UMAPdata, th.size = 50, plot.device = TRUE)
```

## **Arguments**

UMAPdata a UMAPdata object

th. size a numeric value providing the minimum number of cells needed for a cluster to

be considered a small cluster

plot.device a boolean value specifying a results representation must be displayed

# Value

a numerical value corresponding to the percentage of cell cluster with low number of cells

22 QCUniformClusters

QCUniformClusters Computes the percentage of clusters with uniform phenotypes

#### **Description**

This function aims to identify and show cell clusters having a uniform phenotype

A uniform cluster corresponds to a cluster that have a unimodal expression and a low spread of expression for all its markers

# Usage

```
QCUniformClusters(
   UMAPdata,
   uniform.test = c("both", "uniform", "IQR"),
   th.pvalue = 0.05,
   th.IQR = 2,
   plot.device = TRUE
)
```

## **Arguments**

uniform.test a character providing the name of test assessment to perform. Possible value are : 'both', 'uniform', 'IQR'

th.pvalue a numeric value providing the p-value threshold of the Hartigan's dip test (unimodal if pvalue > th.pvalue)

th.IQR a numeric value providing the IQR (interquartile range) threshold to assume a distribution as uniform

plot.device a boolean value specifying if result representation must be displayed

## **Details**

- -'uniform' corresponds to the verification of the unimodal distribution of markers with a Hartigans test
- -'IQR' corresponds to the verification of the distribution of markers so that they are not below the IQR threshold (interquantile range)
- -'both' corresponds to the combination of the two parameters: uniform and IQR

## Value

a numerical value corresponding to the percentage of cell cluster with unimodal expression and a low spread

renameMarkers 23

renameMarkers

Renames markers within a UMAPdata object

## **Description**

This function aims to rename cell markers stored within a UMAPdata object.

This function is interesting to remove the names of the fluorochromes or metals recorded during the acquisition process.

## Usage

```
renameMarkers(UMAPdata, marker.names)
```

# **Arguments**

UMAPdata a UMAPdata object

marker.names a character vector providing the new marker names to use

## Value

a S4 object of class 'UMAPdata'

show

Prints information for a UMAPdata objects

# Description

Shows a preview for a UMAPdata object.

## Usage

```
## S4 method for signature 'UMAPdata'
show(object)
```

## **Arguments**

object

a UMAPdata object

# Value

none

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UMAPdata-class

UMAPdata class definition

## **Description**

The UMAPdata object is a S4 object containing all cytometry expressions.

#### **Slots**

samples a character vector containing the names of the biological samples

raw.markers a character vector containing the names of the raw markers

matrix.expression a data.frame containing the marker expressions of each cell

manifold a data.frame containing the manifold coordinates

manifold.params a list containing the parameters used for manifold creation

recognize.clusters a vector containing the identifiers of cell clusters

recognize.clusters.params a vector containing the parameters used for the identification of the cell clusters

concave.hulls a data.frame containing the coordinates of the cell cluster of the concave hulls for each cluster

matrix.cell.count a data.frame containing the number of cells associated to each cluster for each sample

matrix.abundance a data.frame containing the percentage of cells associated to each cluster for each sample

statistic a data.frame containing the statistics of cell clusters

metadata a data.frame containing the metadata associated to each sample

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