

Package ‘scReClassify’

June 7, 2023

Type Package

Title scReClassify: post hoc cell type classification of single-cell RNA-seq data

Version 1.6.0

Description A post hoc cell type classification tool to fine-tune cell type annotations generated by any cell type classification procedure with semi-supervised learning algorithm AdaSampling technique. The current version of scReClassify supports Support Vector Machine and Random Forest as a base classifier.

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BugReports <https://github.com/SydneyBioX/scReClassify/issues>

URL <https://github.com/SydneyBioX/scReClassify>,
<http://www.bioconductor.org/packages/release/bioc/html/scReClassify.html>

Depends R (>= 4.1)

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SingleCellExperiment, methods

VignetteBuilder knitr

biocViews Software, Transcriptomics, SingleCell, Classification,
SupportVectorMachine

Suggests testthat, knitr, BiocStyle, rmarkdown, DT, mclust, dplyr

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bAccuracy	<i>bAccuracy</i>
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Description

This function calculates the accuracy of the prediction to the true label.

Usage

bAccuracy(cls.truth, final)

Arguments

- cls.truth A character vector of true class label.
- final A vector of final classified label prediction from multiAdaSampling.

Value

An accuracy value.

Author(s)

Pengyi Yang, Taiyun Kim

Examples

```
data("gse87795_subset_sce")

mat.expr <- gse87795_subset_sce
cellTypes <- gse87795_subset_sce$cellTypes

# Get dimension reduced matrix. We are using `logNorm` assay from `mat.expr`.
mat.pc <- matPCs(mat.expr, assay = "logNorm")
```

```
# Here we are using Support Vector Machine as a base classifier.
result <- multiAdaSampling(mat.pc, cellTypes, classifier = "svm",
percent = 1, L = 10)

final <- result$final

# Balanced accuracy
bacc <- bAccuracy(cellTypes, final)
```

gse87795_subset_sce	<i>GSE827795 subset data</i>
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Description

A SingleCellExperiment object containing a subset expression matrix of GSE827795. The data contains log2 transformed FPKM expression.

GSE87795 is a mouse fetal liver development data containing 1000 genes, 367 cells and 6 cell types.

The original GSE87795 data and the study details can be found at this [link](#)

Usage

```
gse87795_subset_sce
```

Format

An object of class SingleCellExperiment with 1000 rows and 367 columns.

matPCs	<i>matPCs function</i>
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Description

Performs PCA on a given matrix and returns a dimension reduced matrix which captures at least 80% (default) of overall variability.

Usage

```
matPCs(data, assay = NULL, percentVar = 0.8)
```

Arguments

data	An expression matrix or a SingleCellExperiment object.
assay	An assay to select if data is a SingleCellExperiment object
percentVar	The percentage of variance threshold. This is used to select number of Principal Components.

Details

This function performs PCA to reduce the dimension of the gene expression matrix limited from 10 to 20 PCs.

Value

Dimensionally reduced matrix.

Author(s)

Pengyi Yang, Taiyun Kim

Examples

```
data("gse87795_subset_sce")

mat.expr <- gse87795_subset_sce

mat.pc <- matPCs(mat.expr, assay = "logNorm")

# to capture at least 70% of overall variability in the dataset,
mat.dim.reduct.70 <- matPCs(mat.expr, assay = "logNorm", 0.7)
```

multiAdaSampling

multi Adaptive Sampling function

Description

Performs multiple adaptive sampling to train a classifier model.

Usage

```
multiAdaSampling(
  data,
  label,
  reducedDimName = NULL,
  classifier = "svm",
  percent = 1,
  L = 10,
  prob = FALSE,
  balance = TRUE,
  iter = 3
)
```

Arguments

<code>data</code>	A dimension reduced matrix from <code>matPCs</code> .
<code>label</code>	A named vector of label information for each sample. The names should match the sample names of <code>data</code>
<code>reducedDimName</code>	A name of the <code>reducedDim</code> to use. This must be specified if <code>data</code> is a <code>SingleCellExperiment</code> object.
<code>classifier</code>	Base classifier model, either "SVM" (<code>svm</code>) or "RF" ' <code>rf</code> ' is supported.
<code>percent</code>	Percentage of samples to select at each iteration.
<code>L</code>	Number of ensembles. Default to 10.
<code>prob</code>	logical flag to return sample's probabilities to each class.
<code>balance</code>	logical flag to if the cell types are balanced. If <code>FALSE</code> , down sample large cell types classes to the median of all class sizes.
<code>iter</code>	A number of iterations to perform <code>adaSampling</code> .

Value

A final prediction, probabilities for each cell type and the model are returned as a list.

Author(s)

Pengyi Yang, Taiyun Kim

Examples

```
library(SingleCellExperiment)

# Loading the data
data("gse87795_subset_sce")

mat.expr <- gse87795_subset_sce
cellTypes <- gse87795_subset_sce$cellTypes

# Get dimension reduced matrix. We are using `logNorm` assay from `mat.expr`.
reducedDim(mat.expr, "matPCs") <- matPCs(mat.expr, assay = "logNorm")

# Here we are using Support Vector Machine as a base classifier.
result <- multiAdaSampling(mat.expr, cellTypes, reducedDimName = "matPCs",
  classifier = "svm", percent = 1, L = 10)
```

scReClassify

scReClassify: a package for post hoc cell type classification of single-cell RNA-sequencing data.

Description

A post hoc cell type classification tool to fine-tune cell type annotations generated by any cell type classification procedure with semi-supervised learning algorithm AdaSampling technique.

The current version of scReClassify supports Support Vector Machine and Random Forest as a base classifier.

Author(s)

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See Also

Useful links:

- Vignette available at: <https://sydneybio.github.io/scdney/>

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