

Package ‘GloScope’

December 10, 2025

Type Package

Title Population-level Representation on scRNA-Seq data

Version 2.1.2

Description This package aims at representing and summarizing the entire single-cell profile of a sample. It allows researchers to perform important bioinformatic analyses at the sample-level such as visualization and quality control. The main functions Estimate sample distribution and calculate statistical divergence among samples, and visualize the distance matrix through MDS plots.

BugReports <https://github.com/epurdom/GloScope/issues>

License Artistic-2.0

Encoding UTF-8

Imports utils, stats, MASS, mclust, ggplot2, RANN, FNN, BiocParallel, mvnfast, SingleCellExperiment, rlang, RColorBrewer, pheatmap, vegan, cluster, boot, permute

Depends R (>= 4.4.0)

Suggests BiocStyle, testthat (>= 3.0.0), knitr, rmarkdown, zellkonverter

VignetteBuilder knitr

LazyData false

biocViews DataRepresentation, QualityControl, RNASeq, Sequencing, Software, SingleCell

RoxygenNote 7.3.3

Config/testthat/edition 3

git_url <https://git.bioconductor.org/packages/GloScope>

git_branch devel

git_last_commit 458539a

git_last_commit_date 2025-11-17

Repository Bioconductor 3.23

Date/Publication 2025-12-09

Author Elizabeth Purdom [aut, cre],
William Torous [aut] (ORCID: <<https://orcid.org/0000-0001-5668-5510>>),
Hao Wang [aut] (ORCID: <<https://orcid.org/0000-0002-0749-474X>>),
Boying Gong [aut]

Maintainer Elizabeth Purdom <epurdom@stat.berkeley.edu>

Contents

example_SCE	2
getMetrics	3
gloscope	6
gloscopeProp	8
plotCI	9
plotHeatmap	10
plotMDS	11
Index	13

example_SCE	<i>SingleCellExperiment containing example inputs to GloScope</i>
-------------	---

Description

‘example_SCE’ is a SingleCellExperiment object which contains PCA embeddings and metadata for PBMCs from 20 COVID-infected and healthy control patients. Each sample is reduced to a random subset of 500 cells, for a total of 10,000 cells. The ‘colData’ slot of the object contains the metadata for each cell, its sample ID and phenotype. The dimensionality reductions slot contains the first 50 PCs, and these embeddings are provided by the authors of "Single-cell multi-omics analysis of the immune response in COVID-19" (Stephenson et al., 2021; doi: 10.1038/s41591-021-01329-2).

‘example_SCE_small’ is a SingleCellExperiment with the same structure as ‘example_SCE’, but only containing data from the first five samples. This is a smaller set for examples.

Format

A SingleCellExperiment object with metadata and PCA embeddings

Value

A SingleCellExperiment object

Examples

```
# Code to create the small SCE from the full sample
# Reduction to 5 samples demonstrates data extraction from SCE objects
data(example_SCE)
sample_ids <- SingleCellExperiment::colData(example_SCE)$sample_id
whKeep <- which(sample_ids %in% unique(sample_ids)[seq_len(5)])
example_SCE_small <- SingleCellExperiment::SingleCellExperiment(
  assays=list(counts=matrix(rep(0,2500),ncol=2500)),
  colData=SingleCellExperiment::colData(example_SCE)[whKeep,],
  reducedDims=list("PCA"=SingleCellExperiment::reducedDim(example_SCE,"PCA")[whKeep,]))
```

getMetrics	<i>Calculate metrics on separation in distance matrix due to grouping variable</i>
------------	--

Description

These functions are wrappers for calculating common metrics for the amount of separation in a distance matrix due to a grouping (factor) variable and creating bootstrap confidence intervals and permutation tests.

Usage

```
getMetrics(
  dist_mat,
  metadata_df,
  metrics = c("anosim", "adonis2", "silhouette"),
  sample_id,
  group_vars,
  checkData = TRUE,
  permuteTest = FALSE,
  permutations = 100
)

bootCI(
  dist_mat,
  metadata_df,
  metrics = "anosim",
  sample_id,
  group_vars,
  R = 1000,
  ci_type = c("perc", "norm", "basic", "stud", "bca"),
  ci_conf = 0.95,
  ...
)

bootGloscope(
```

```

    dist_mat,
    metadata_df,
    metrics = "anosim",
    sample_id,
    group_vars,
    R = 1000,
    ...
)

```

Arguments

<code>dist_mat</code>	The divergence matrix output of <code>'gloscope()'</code> . Should be a symmetric, square matrix. For <code>'bootCI'</code> the argument can be a list of distance matrices.
<code>metadata_df</code>	A data frame contains each sample's metadata. Note this is NOT at the cell-level, and should have the same number of rows as <code>dist_mat</code> .
<code>metrics</code>	vector of statistics to calculate. For <code>'bootstrap_gloscope'</code> must be single value.
<code>sample_id</code>	The column name or index in <code>metadata_df</code> that contains the sample ID. This is for ensuring alignment between the <code>dist_mat</code> and the <code>metadata_df</code> . The row-names of <code>dist_mat</code> are expected to match the <code>sample_id</code> values.
<code>group_vars</code>	vector of names of grouping variables in <code>metadata_df</code> for which to calculate metrics. For <code>'bootstrap_gloscope'</code> must be single value.
<code>checkData</code>	Whether to check whether <code>dist_mat</code> , <code>metadata_df</code> , and <code>sample_id</code> match, for example in terms of dimensions and rownames. Mainly used internally.
<code>permuteTest</code>	whether to run permutation tests on each of the metrics
<code>permutations</code>	if <code>'permuteTest=TRUE'</code> , an integer value defines the number of permutations. Can also except output of how for more fine control of the permutation mechanisms.
<code>R</code>	number of bootstrap replicates. See boot .
<code>ci_type</code>	Single character value. The type of confidence interval to compute. Passed to argument <code>'type'</code> in boot.ci
<code>ci_conf</code>	Scalar value between 0 and 1. The confidence level requested. Passed to argument <code>'conf'</code> in boot.ci .
<code>...</code>	arguments passed to boot

Details

The function `'getMetrics'` is a simple wrapper for calculating statistics that summarize the difference between distances within and between groupings. If the variable defined by `group_var` does not have at least two groupings, the function will return a NA.

The options "anosim" and "adonis2" are wrappers to the functions of that name in the package `'vegan'`; we have turned off the permutation testing option of those functions. The functions in `'vegan'` have greater capability, and in particular [adonis2](#) has capability to handle more complicated testing paradigms than a simple grouping factor. Permutation tests for these statistics are handled by the functions of `'vegan'`.

The option "silhouette" calls the function of that name from the package 'cluster' and calculates the silhouette width of each group and then averages them across groups. The permutation test is coded making use of the package 'permute', similar to 'vegan', so that control of the permutation mechanism is possible in the same way.

'bootCI' is a wrapper function to [boot.ci](#). 'boot.ci' can be called directly on the output of 'bootGloscope'. The main advantage of 'bootCI' is to calculate bootstrap CI over multiple choices of metrics, variables, and/or distance matrices. Unlike 'boot.ci', 'bootCI' does not allow different choices of confidence interval types or levels, so 'ci_type' and 'ci_level' must be of length 1. For this kind of multiplicity, call 'boot.ci' directly on the output of 'bootGloscope'.

The function 'bootGloscope' is a wrapper to the [boot](#) function for creating bootstraps of one of the metrics calculated by 'getMetrics'. Most users will probably prefer 'bootCI'

Value

'getMetrics' creates a data frame containing the statistic for each combination of metric and grouping variable with columns

- metric
- grouping
- statistic
- pval (if 'permuteTest=TRUE')

'bootCI' creates a data frame containing the statistic for each combination of metric and grouping variable with columns with the upper and lower bounds of the requested confidence intervals

- metric
- grouping
- statistic
- lower
- upper

'bootGloscope' returns an object of class 'boot' created by [boot](#).

See Also

[anosim](#), [adonis2](#), [silhouette](#), [how](#)

[boot.ci](#)

[boot](#)

Examples

```
data(example_SCE_small)
sample_ids <- SingleCellExperiment::colData(example_SCE_small)$sample_id
# Run gloscope on first 10 PCA embeddings
# We use 'KNN' option for speed ('GMM' is slightly slower)
pca_embeddings <- SingleCellExperiment::reducedDim(example_SCE_small, "PCA")
pca_embeddings_subset <- pca_embeddings[,seq_len(10)] # select the first 10 PCs
dist_result <- gloscope(pca_embeddings_subset, sample_ids,
```

```

    dens="KNN",
    BPPARAM = BiocParallel::SerialParam(RNGseed=2))
# make a per-sample metadata
sample_metadata <- as.data.frame(unique(SingleCellExperiment::colData(example_SCE_small)[,c(1,2)]))
# make another variable
sample_metadata$grouping<-c(rep(c("A", "B"),each=2),"A")
getMetrics(dist_result,metadata_df=sample_metadata, sample_id="sample_id",
  group_vars="phenotype")
# run permutation tests:
getMetrics(dist_result,metadata_df=sample_metadata, sample_id="sample_id",
  group_vars=c("phenotype","grouping"), permuteTest=TRUE)

# calculate many bootstraps -- for speed up we set R ridiculously low
manyboot<-bootCI(list("Distance 1"=dist_result,"Another distance"=dist_result),
  sample_metadata,"sample_id",
  metrics=c("anosim","silhouette"),group_vars=c("phenotype","grouping"),R=20)

# single bootstrap of anosim
bootout<-bootGloscope(dist_result,sample_metadata,"sample_id",
  metric="anosim",group_var="phenotype")
#work with the boot object using functions in boot package:
library(boot)
print(bootout)
boot.ci(bootout)

```

gloscope

Calculate statistical divergence between all sample pairs

Description

This function calculates a matrix of pairwise divergences between input samples of single cell data.

Usage

```

gloscope(
  embedding_matrix,
  cell_sample_ids,
  dens = c("GMM", "KNN"),
  dist_metric = c("KL", "JS"),
  r = 10000,
  num_components = c(5, 10, 15, 20),
  k = 50,
  GMM_params = list(modelNames = c("VVE"), verbose = FALSE, plot = FALSE),
  KNN_params = NULL,
  BPPARAM = BiocParallel::SerialParam(),
  prefit_density = NULL,
  return_density = FALSE
)

```

Arguments

<code>embedding_matrix</code>	a matrix or data.frame of latent embeddings with rows corresponding to cells and columns to dimensions
<code>cell_sample_ids</code>	a vector of the samples IDs each cell comes from. Length must match the number of rows in <code>'embedding_matrix'</code>
<code>dens</code>	the density estimation. One of <code>c("GMM", "KNN")</code>
<code>dist_metric</code>	distance metric to calculate the distance. One of <code>c("KL", "JS")</code>
<code>r</code>	number of Monte Carlo simulations to generate
<code>num_components</code>	a vector of integers for the number of components to fit GMMs to, default is <code>c(5,10,15,20)</code>
<code>k</code>	number of nearest neighbours for KNN density estimation, default <code>k = 50</code> .
<code>GMM_params</code>	optional mclust parameters, default is to restrict the fit model to only VVE
<code>KNN_params</code>	optional arguments for either <code>'FNN:KL.dist'</code> (KL) or <code>'RANN::nn2'</code> (JS), default is NULL
<code>BPPARAM</code>	BiocParallel parameters, default is running in serial. Set random seed with <code>'RNGseed'</code> argument
<code>prefit_density</code>	a named list of pre-fit <code>'densityMclust'</code> objects for each sample, default is NULL
<code>return_density</code>	return the GMM parameter list or not (if applicable), default is FALSE

Value

A matrix containing the pairwise divergence or distance between all pairs of samples

Examples

```
# Bring in small example data of single cell embeddings
data(example_SCE_small)
sample_ids <- SingleCellExperiment::colData(example_SCE_small)$sample_id
pca_embeddings <- SingleCellExperiment::reducedDim(example_SCE_small, "PCA")
# Run gloscope on first 10 PCA embeddings
# We use 'KNN' option for speed ('GMM' is slightly slower)
pca_embeddings_subset <- pca_embeddings[,seq_len(10)] # select the first 10 PCs
dist_result <- gloscope(pca_embeddings_subset, sample_ids,
  dens="KNN", BPPARAM = BiocParallel::SerialParam(RNGseed=2))
dist_result
```

plotCI	<i>Plot confidence intervals</i>
--------	----------------------------------

Description

This function creates a ‘ggplot’ object that plots the confidence intervals created by ‘bootCI’

Usage

```
plotCI(ci_df, color_by, group_by, dodge_width = 0.5)
```

Arguments

ci_df	A data frame contains each sample’s metadata. Note this is NOT at the cell-level.
color_by	The column name or index in ci_df that should be used to color the confidence intervals by.
group_by	The column name or index in ci_df that should be used to determine how to group the confidence intervals. If missing all confidence intervals will be plotted in an order determined internally.
dodge_width	value passed to ‘width’ argument of position_dodge if ‘group_by’ variable is given. Controls separation between confidence intervals with the same grouping value.

Value

A plot of sample-pair divergences with confidence intervals

Examples

```
data(example_SCE_small)
sample_ids <- SingleCellExperiment::colData(example_SCE_small)$sample_id
# Run gloscope on first 10 PCA embeddings
# We use 'KNN' option for speed ('GMM' is slightly slower)
pca_embeddings <- SingleCellExperiment::reducedDim(example_SCE_small,"PCA")
pca_embeddings_subset <- pca_embeddings[,seq_len(10)] # select the first 10 PCs
dist_result <- gloscope(pca_embeddings_subset, sample_ids,
  dens="KNN",
  BPPARAM = BiocParallel::SerialParam(RNGseed=2))
# make a per-sample metadata
sample_metadata <- as.data.frame(unique(SingleCellExperiment::colData(example_SCE_small)[,c(1,2)]))
# make another variable
sample_metadata$fakeGroup<-c(rep(c("A","B"),each=2),"A")
manyboot<-bootCI(dist_result,
  sample_metadata,"sample_id",
  metrics=c("anosim","silhouette"),group_vars=c("phenotype","fakeGroup"),R=20)
plotCI(manyboot,group_by="metric",color_by="grouping")
```

plotHeatmap

Plot a heatmap of the GloScope representation

Description

This function creates a heatmap of the given GloScope divergence matrix.

Usage

```
plotHeatmap(
  dist_mat,
  metadata_df,
  sample_id,
  color_by,
  which_side = c("columns", "rows", "both"),
  ...
)
```

Arguments

dist_mat	The divergence matrix output of ‘gloscope()’. Should be a symmetric, square matrix.
metadata_df	A data frame contains each sample’s metadata. Note this is NOT at the cell-level, and should have the same number of rows as dist_mat.
sample_id	The column name or index in metadata_df that contains the sample ID. This is for ensuring alignment between the dist_mat and the metadata_df. The row-names of dist_mat are expected to match the sample_id values.
color_by	A vector of column names or indices in metadata_df that should be used to color/annotate the samples.
which_side	One of "columns", "rows", or "both", indicating whether the annotation of the samples in ‘color_by’ should be on the rows, columns, or on both.
...	parameters passed to pheatmap .

Details

The function is a wrapper function to [pheatmap](#). ‘color_by’ is used to create subset of the ‘metadata_df’ to pass to ‘annotation_col’ (if ‘which_side="columns"’) or ‘annotation_row’ (if ‘which_side="rows"’). If ‘which_side="both"’, then it is passed to both, and ‘annotation_names_row’ argument is set to ‘FALSE’, suppressing labeling both the columns and rows (which user can thus not override). All other arguments to [pheatmap](#) can be passed directly by the user

Value

Invisibly returns the output of [pheatmap](#)

See Also[pheatmap](#)**Examples**

```

data(example_SCE_small)
sample_ids <- SingleCellExperiment::colData(example_SCE_small)$sample_id
# Run gloscope on first 10 PCA embeddings
# We use 'KNN' option for speed ('GMM' is slightly slower)
pca_embeddings <- SingleCellExperiment::reducedDim(example_SCE_small,"PCA")
pca_embeddings_subset <- pca_embeddings[,seq_len(10)] # select the first 10 PCs
dist_result <- gloscope(pca_embeddings_subset, sample_ids,
  dens="KNN",
  BPPARAM = BiocParallel::SerialParam(RNGseed=2))
# make a per-sample metadata
sample_metadata <- as.data.frame(unique(SingleCellExperiment::colData(example_SCE_small)[,c(1,2)]))
plotHeatmap(dist_mat = dist_result, metadata_df = sample_metadata ,
  sample_id="sample_id", color_by="phenotype")
# Pass additional options to pheatmap to control colors of groups
library(RColorBrewer)
plotHeatmap(dist_mat = dist_result, metadata_df = sample_metadata ,
  sample_id="sample_id", color_by="phenotype", which_side="both",
  annotation_colors=list(phenotype = c(Covid = "magenta", Healthy = "white")),
  color = colorRampPalette(brewer.pal(9, "PuBuGn"))(100))

```

plotMDS

*Plot the multidimensional scaling of the GloScope representation***Description**

This function calculates the multidimensional scaling for a GloScope divergence matrix and returns a ggplot object that plots it.

Usage

```
plotMDS(dist_mat, metadata_df, sample_id, k = 10, color_by, shape_by)
```

Arguments

dist_mat	The divergence matrix output of 'gloscope()'. Should be a symmetric, square matrix.
metadata_df	A data frame contains each sample's metadata. Note this is NOT at the cell-level, and should have the same number of rows as dist_mat.
sample_id	The column name or index in metadata_df that contains the sample ID. This is for ensuring alignment between the dist_mat and the metadata_df. The row-names of dist_mat are expected to match the sample_id values.
k	Number of MDS dimension to generate, default = 10

color_by	The column name or index in metadata_df that should be used to color the points by. If missing all points will be the same color.
shape_by	The column name or index in metadata_df that should be used to determine the shape of the points. If missing all points will be the same shape.

Details

The function calls [isoMDS](#) from the MASS package, calculates the requested k coordinates of the MDS plot. It also creates a ggplot object that will plot the first two dimensions color or shape coded by the given variables in the metadata data frame.

Value

A list containing the MDS embedding and plot of the distance matrix

- mds - A data.frame containing the MDS embedding, with the number of rows equal to the number of samples.
- plot - A ggplot object containing the plot object. 'print' of the object will create a plot.

See Also

[isoMDS](#)

Examples

```
data(example_SCE_small)
sample_ids <- SingleCellExperiment::colData(example_SCE_small)$sample_id
# Run gloscope on first 10 PCA embeddings
# We use 'KNN' option for speed ('GMM' is slightly slower)
pca_embeddings <- SingleCellExperiment::reducedDim(example_SCE_small,"PCA")
pca_embeddings_subset <- pca_embeddings[,seq_len(10)] # select the first 10 PCs
dist_result <- gloscope(pca_embeddings_subset, sample_ids,
  dens="KNN",
  BPPARAM = BiocParallel::SerialParam(RNGseed=2))
# make a per-sample metadata
sample_metadata <- as.data.frame(unique(SingleCellExperiment::colData(example_SCE_small)[,c(1,2)]))
mds_result <- plotMDS(dist_mat = dist_result, metadata_df = sample_metadata ,
  sample_id="sample_id", color_by="phenotype",k=2)
head(mds_result$mds)
require(ggplot2)
mds_result$plot
# Add additional ggplot2 components to adapt figure
mds_result$plot + theme_bw() + scale_color_manual(values=alpha(c("red","blue"),0.5))
```

Index

adonis2, [4](#), [5](#)

anosim, [5](#)

boot, [4](#), [5](#)

boot.ci, [4](#), [5](#)

bootCI (getMetrics), [3](#)

bootGloscope (getMetrics), [3](#)

example_SCE, [2](#)

example_SCE_small (example_SCE), [2](#)

getMetrics, [3](#)

gloscope, [6](#)

gloscopeProp, [8](#)

how, [4](#), [5](#)

isoMDS, [12](#)

pheatmap, [10](#), [11](#)

plotCI, [9](#)

plotHeatmap, [10](#)

plotMDS, [11](#)

position_dodge, [9](#)

silhouette, [5](#)