Package ‘SC3’

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Author Vladimir Kiselev
Maintainer Vladimir Kiselev <vladimir.yu.kiselev@gmail.com>
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**calculate_distance**

*Calculate a distance matrix*

**Description**

Distance between the cells, i.e. columns, in the input expression matrix are calculated using the Euclidean, Pearson and Spearman metrics to construct distance matrices.

**Usage**

```r
calculate_distance(data, method)
```

**Arguments**

- `data` : expression matrix
- `method` : one of the distance metrics: 'spearman', 'pearson', 'euclidian'
### calculate_stability

**Value**

distance matrix

---

### Description

Calculate the stability index of the obtained clusters when changing k.

Stability index shows how stable each cluster is across the selected range of k. The stability index varies between 0 and 1, where 1 means that the same cluster appears in every solution for different k.

**Usage**

```r
calculate_stability(consensus, k)
```

**Arguments**

- `consensus`: consensus item of the sc3 slot of an object of 'SCESet' class
- `k`: number of clusters

**Details**

Formula (imagine a given cluster with is split into N clusters when k is changed, and in each of the new clusters there are given_cells of the given cluster and also some extra_cells from other clusters): $SI = \frac{\sum_{ks} \sum_{clusters} N(\text{given_cells}/(\text{given_cells} + \text{extra_cells}))}{N(\text{corrects for stability of each cluster})/N(\text{corrects for the number of clusters})/\text{length}(ks)}$

**Value**

a numeric vector containing a stability index of each cluster

---

### consensus_matrix

**Calculate consensus matrix**

**Description**

Consensus matrix is calculated using the Cluster-based Similarity Partitioning Algorithm (CSPA). For each clustering solution a binary similarity matrix is constructed from the corresponding cell labels: if two cells belong to the same cluster, their similarity is 1, otherwise the similarity is 0. A consensus matrix is calculated by averaging all similarity matrices.

**Usage**

```r
consensus_matrix(clusts)
```

**Arguments**

- `clusts`: a matrix containing clustering solutions in columns
Value

consensus matrix

---

consmx  Consensus matrix computation

Description

Computes consensus matrix given cluster labels

Usage

consmx(dat)

Arguments

dat  a matrix containing clustering solutions in columns

---

ED1  Compute Euclidean distance matrix by rows

Description

Used in consmx function

Usage

ED1(x)

Arguments

x  A numeric matrix.

---

ED2  Compute Euclidean distance matrix by columns

Description

Used in sc3-funcs.R distance matrix calculation and within the consensus clustering.

Usage

ED2(x)

Arguments

x  A numeric matrix.
**Estimate the optimal k for k-means clustering**

**Description**

The function finds the eigenvalues of the sample covariance matrix. It will then return the number of significant eigenvalues according to the Tracy-Widom test.

**Usage**

```r
estkTW(dataset)
```

**Arguments**

- `dataset` processed input expression matrix.

**Value**

an estimated number of clusters k

---

**Calculate the area under the ROC curve for a given gene.**

**Description**

For a given gene a binary classifier is constructed based on the mean cluster expression values (these are calculated using the cell labels). The classifier prediction is then calculated using the gene expression ranks. The area under the receiver operating characteristic (ROC) curve is used to quantify the accuracy of the prediction. A p-value is assigned to each gene by using the Wilcoxon signed rank test.

**Usage**

```r
get_auroc(gene, labels)
```

**Arguments**

- `gene` expression data of a given gene
- `labels` cell labels corresponding to the expression values of the gene
get_biolgy  

Wrapper for calculating biological properties

Description
Wrapper for calculating biological properties

Usage
get_biolgy(dataset, labels, regime)

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>dataset</td>
<td>expression matrix</td>
</tr>
<tr>
<td>labels</td>
<td>cell labels corresponding clusters</td>
</tr>
<tr>
<td>regime</td>
<td>defines what biological analysis to perform. &quot;marker&quot; for marker genes, &quot;de&quot; for differentially expressed genes and &quot;outl&quot; for outlier cells</td>
</tr>
</tbody>
</table>

Value
results of either

get_de_genes  

Find differentially expressed genes

Description
Differential expression is calculated using the non-parametric Kruskal-Wallis test. A significant p-value indicates that gene expression in at least one cluster stochastically dominates one other cluster. Note that the calculation of differential expression after clustering can introduce a bias in the distribution of p-values, and thus we advise to use the p-values for ranking the genes only.

Usage
get_de_genes(dataset, labels)

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>dataset</td>
<td>expression matrix</td>
</tr>
<tr>
<td>labels</td>
<td>cell labels corresponding to the columns of the expression matrix</td>
</tr>
</tbody>
</table>

Value
a numeric vector containing the differentially expressed genes and corresponding p-values

Examples

d <- get_de_genes(treutlein[1:10, ], colnames(treutlein))
head(d)
**get_marker_genes**

*Calculate marker genes*

**Description**

Find marker genes in the dataset. The `get_auroc` is used to calculate marker values for each gene.

**Usage**

```r
get_marker_genes(dataset, labels)
```

**Arguments**

- **dataset**
  - expression matrix
- **labels**
  - cell labels corresponding clusters

**Value**

data.frame containing the marker genes, corresponding cluster indexes and adjusted p-values

**Examples**

```r
d <- get_marker_genes(treutlein[1:10,], colnames(treutlein))
d
```

---

**get_outl_cells**

*Find cell outliers in each cluster.*

**Description**

Outlier cells in each cluster are detected using robust distances, calculated using the minimum covariance determinant (MCD), namely using `covMcd`. The outlier score shows how different a cell is from all other cells in the cluster and it is defined as the differences between the square root of the robust distance and the square root of the 99.99

**Usage**

```r
get_outl_cells(dataset, labels)
```

**Arguments**

- **dataset**
  - expression matrix
- **labels**
  - cell labels corresponding to the columns of the expression matrix

**Value**

a numeric vector containing the cell labels and corresponding outlier scores ordered by the labels
**Examples**

d <- get_outl_cells(treutlein[1:10,, colnames(treutlein)])
head(d)

**get_processed_dataset**  
*Get processed dataset used by SC3 from the default scater slots*

**Description**
Takes data from the 'exprs' slot and applies the gene filter

**Usage**
get_processed_dataset(object)

**Arguments**

- **object**
  - an object of 'SCESet' class

**markers_for_heatmap**  
*Reorder and subset gene markers for plotting on a heatmap*

**Description**
Reorders the rows of the input data.frame based on the `sc3_k_markers_clusts` column and also keeps only the top 10 genes for each value of `sc3_k_markers_clusts`.

**Usage**
markers_for_heatmap(markers)

**Arguments**

- **markers**
  - a data.frame object with the following colnames: `sc3_k_markers_clusts, sc3_k_markers_auroc, sc3_k_markers_padj`

**norm_laplacian**  
*Graph Laplacian calculation*

**Description**
Calculate graph Laplacian of a symmetrix matrix

**Usage**
norm_laplacian(A)

**Arguments**

- **A**
  - symmetric matrix
organise_de_genes

Get differentially expressed genes from an object of SCESet class

Description
This function returns all marker gene columns from the phenoData slot of the input object corresponding to the number of clusters $k$. Additionally, it rearranges genes by the cluster index and order them by the area under the ROC curve value inside of each cluster.

Usage
organise_de_genes(object, k, p_val)

Arguments
- object: an object of SCESet class
- k: number of cluster
- p_val: p-value threshold

organise_marker_genes
Get marker genes from an object of SCESet class

Description
This function returns all marker gene columns from the phenoData slot of the input object corresponding to the number of clusters $k$. Additionally, it rearranges genes by the cluster index and order them by the area under the ROC curve value inside of each cluster.

Usage
organise_marker_genes(object, k, p_val, auroc)

Arguments
- object: an object of SCESet class
- k: number of cluster
- p_val: p-value threshold
- auroc: area under the ROC curve threshold
**prepare_for_svm**  
* A helper function for the SVM analysis

**Description**

 Defines train and study cell indeces based on the svm_num_cells and svm_train_inds input parameters

**Usage**

 `prepare_for_svm(N, svm_num_cells = NULL, svm_train_inds = NULL, svm_max)`

**Arguments**

- **N**
  - number of cells in the input dataset
- **svm_num_cells**
  - number of random cells to be used for training
- **svm_train_inds**
  - indeces of cells to be used for training
- **svm_max**
  - define the maximum number of cells below which SVM is not run

**Value**

 A list of indeces of the train and the study cells

---

**reindex_clusters**  
* Reindex cluster labels in ascending order

**Description**

 Given an `hclust` object and the number of clusters k this function reindex the clusters inferred by `cutree(hc, k)[hc$order]`, so that they appear in ascending order. This is particularly useful when plotting heatmaps in which the clusters should be numbered from left to right.

**Usage**

 `reindex_clusters(hc, k)`

**Arguments**

- **hc**
  - an object of class hclust
- **k**
  - number of cluster to be inferred from hc

**Examples**

 `hc <- hclust(dist(USArrests), 'ave')`  
 `cutree(hc, 10)[hc$order]`  
 `reindex_clusters(hc, 10)[hc$order]`
Run all steps of SC3 in one go

Description

This function is a wrapper that executes all steps of SC3 analysis in one go. Please note that by default the "exprs" slot of the input scater object is used for the SC3 analysis. If the scater object has been created in a standard way then the expression values in the "exprs" slot will be automatically log-transformed. If you have overwritten the "exprs" slot manually, please make sure that the values in the "exprs" slot are log-transformed before running the SC3 analysis. SC3 assumes that the data is log-transformed by default.

Usage

```r
sc3.SCESet(object, ks = NULL, gene_filter = TRUE, pct_dropout_min = 10,
pct_dropout_max = 90, d_region_min = 0.04, d_region_max = 0.07,
svm_num_cells = NULL, svm_train_inds = NULL, svm_max = 5000,
n_cores = NULL, kmeans_nstart = NULL, kmeans_iter_max = 1e+09,
k_estimator = FALSE, biology = FALSE, rand_seed = 1)
```

```r
## S4 method for signature 'SCESet'
sc3(object, ks = NULL, gene_filter = TRUE,
pct_dropout_min = 10, pct_dropout_max = 90, d_region_min = 0.04,
d_region_max = 0.07, svm_num_cells = NULL, svm_train_inds = NULL,
svm_max = 5000, n_cores = NULL, kmeans_nstart = NULL,
kmeans_iter_max = 1e+09, k_estimator = FALSE, biology = FALSE,
rand_seed = 1)
```

Arguments

- `object`: an object of SCESet class.
- `ks`: a range of the number of clusters \( k \) used for SC3 clustering. Can also be a single integer.
- `gene_filter`: a boolean variable which defines whether to perform gene filtering before SC3 clustering.
- `pct_dropout_min`: if `gene_filter = TRUE`, then genes with percent of dropouts smaller than `pct_dropout_min` are filtered out before clustering.
- `pct_dropout_max`: if `gene_filter = TRUE`, then genes with percent of dropouts larger than `pct_dropout_max` are filtered out before clustering.
- `d_region_min`: defines the minimum number of eigenvectors used for kmeans clustering as a fraction of the total number of cells. Default is 0.04. See SC3 paper for more details.
- `d_region_max`: defines the maximum number of eigenvectors used for kmeans clustering as a fraction of the total number of cells. Default is 0.07. See SC3 paper for more details.
- `svm_num_cells`: number of randomly selected training cells to be used for SVM prediction. The default is NULL.
**sc3_calc_biology**

Calculate DE genes, marker genes and cell outliers.

**Description**

This function calculates differentially expressed (DE) genes, marker genes and cell outliers based on the consensus SC3 clusterings.

**Usage**

```r
sc3_calc_biology(object, ks = NULL, regime = NULL)
```

### S4 method for signature 'SCESet'

```r
sc3_calc_biology.SCESet(object, ks = NULL, regime = NULL)
```

**Arguments**

- **object**: an object of 'SCESet' class
- **ks**: number of clusters k (should be used in the case when a user would like to run k-means on a manually chosen k)
- **regime**: defines what biological analysis to perform. "marker" for marker genes, "de" for differential expressed genes and "outl" for outlier cells
- ... further arguments passed to `sc3_calc_biology.SCESet`

**Value**

an object of SCESet class
Details

DE genes are calculated using `get_de_genes`. Results of the DE analysis are saved as new columns in the `featureData` slot of the input object. The column names correspond to the adjusted p-values of the genes and have the following format: `sc3_k_de_padj`, where k is the number of clusters.

Marker genes are calculated using `get_marker_genes`. Results of the marker gene analysis are saved as three new columns (for each k) to the `featureData` slot of the input object. The column names correspond to the SC3 cluster labels, to the adjusted p-values of the genes and to the area under the ROC curve and have the following format: `sc3_k_markers_clusts`, `sc3_k_markers_padj` and `sc3_k_markers_auroc`, where k is the number of clusters.

Outlier cells are calculated using `get_outl_cells`. Results of the cell outlier analysis are saved as new columns in the `phenoData` slot of the input object. The column names correspond to the log2(outlier_score) and have the following format: `sc3_k_log2_outlier_score`, where k is the number of clusters.

Additionally, biology item is added to the sc3 slot and is set to `TRUE` indicating that the biological analysis of the dataset has been performed.

Value

an object of `SCESet` class
sc3_calc_dists  
*Calculate distances between the cells.*

**Description**

This function calculates distances between the cells contained in the processed_dataset item of the sc3 slot of the SCESet object. It then creates and populates the following items of the sc3 slot:

- **distances** - contains a list of distance matrices corresponding to Euclidean, Pearson and Spearman distances.

Please note that by default the "exprs" slot of the input scater object is used for the SC3 analysis. If the scater object has been created in a standard way then the expression values in the "exprs" slot will be automatically log-transformed. If you have overwritten the "exprs" slot manually, please make sure that the values in the "exprs" slot are log-transformed before running the SC3 analysis. SC3 assumes that the data is log-transformed by default.

**Usage**

sc3_calc_dists.SCESet(object)

```r
## S4 method for signature 'SCESet'
sc3_calc_dists(object)
```

**Arguments**

- **object**  
an object of 'SCESet' class

**Value**

an object of 'SCESet' class

---

sc3_calc_transfs  
*Calculate transformations of the distance matrices.*

**Description**

This function transforms all distances items of the sc3 slot of the SCESet object using either principal component analysis (PCA) or by calculating the eigenvectors of the associated graph Laplacian. The columns of the resulting matrices are then sorted in descending order by their corresponding eigenvalues. The first \( d \) columns (where \( d = \max(\text{object@sc3$n_dim}) \)) of each transformation are then written to the transformations item of the sc3 slot. Additionally, this function also removes the previously calculated distances from the sc3 slot, as they are not needed for further analysis.

**Usage**

sc3_calc_transfs.SCESet(object)

```r
## S4 method for signature 'SCESet'
sc3_calc_transfs(object)
```
sc3_estimate_k

**Arguments**

object  
an object of 'SCESet' class

**Value**

an object of 'SCESet' class

---

### Description

Uses Tracy-Widom theory on random matrices to estimate the optimal number of clusters $k$. Using the function `estkTW` to perform the estimation. It creates and populates the following items of the 'sc3' slot:

- `k_estimation` - contains the estimated value of `$k$`.

Please note that by default the "exprs" slot of the input `scater` object is used for the SC3 analysis. If the `scater` object has been created in a standard way then the expression values in the "exprs" slot will be automatically log-transformed. If you have overwritten the "exprs" slot manually, please make sure that the values in the "exprs" slot are log-transformed before running the SC3 analysis. SC3 assumes that the data is log-transformed by default.

### Usage

```r
sc3_estimate_k.SCESet(object)
```

## S4 method for signature 'SCESet'

```r
sc3_estimate_k(object)
```

**Arguments**

object  
an object of SCESet class

**Value**

an estimated value of $k$
sc3_export_results_xls

Write SC3 results to Excel file

Description

This function writes all SC3 results to an Excel file.

Usage

sc3_export_results_xls.SCESet(object, filename = "sc3_results.xls")

## S4 method for signature 'SCESet'
sc3_export_results_xls(object,
    filename = "sc3_results.xls")

Arguments

- object: an object of 'SCESet' class
- filename: name of the Excel file, to which the results will be written

sc3_interactive

Opens SC3 results in an interactive session in a web browser.

Description

Runs interactive shiny session of SC3 based on precomputed clusterings.

Usage

sc3_interactive.SCESet(object)

## S4 method for signature 'SCESet'
sc3_interactive(object)

Arguments

- object: an object of SCESet class

Value

Opens a browser window with an interactive shiny app and visualize all precomputed clusterings.
**sc3_kmeans**

**kmeans clustering of cells.**

**Description**

This function performs kmeans clustering of the matrices contained in the transformations item of the sc3 slot of the SCESet object. It then creates and populates the following items of the sc3 slot:

- kmeans - contains a list of kmeans clusterings.

**Usage**

```r
sc3_kmeans.SCESet(object, ks = NULL)
## S4 method for signature 'SCESet'
sc3_kmeans(object, ks = NULL)
```

**Arguments**

- `object` - an object of `SCESet` class
- `ks` - number of clusters k (should be used in the case when a user would like to run k-means on a manually chosen k)
- `...` - further arguments passed to `sc3_kmeans.SCESet`

**Details**

See `sc3_prepare` for the default clustering parameters.

**Value**

an object of `SCESet` class

---

**sc3_plot_cluster_stability**

*Plot stability of the clusters*

**Description**

Stability index shows how stable each cluster is across the selected range of ks. The stability index varies between 0 and 1, where 1 means that the same cluster appears in every solution for different k.

**Usage**

```r
sc3_plot_cluster_stability.SCESet(object, k)
## S4 method for signature 'SCESet'
sc3_plot_cluster_stability(object, k)
```
**sc3_plot_de_genes**

**Arguments**

- **object**: an object of 'SCESet' class
- **k**: number of clusters

**Description**

SC3 plots gene expression profiles of the 50 genes with the lowest p-values.

**Usage**

```r
sc3_plot_de_genes.SCESet(object, k, p.val = 0.01, show_pdata = NULL)
```

## S4 method for signature 'SCESet'

```r
sc3_plot_de_genes(object, k, p.val = 0.01, show_pdata = NULL)
```

**sc3_plot_consensus**

*Plot consensus matrix as a heatmap*

**Description**

The consensus matrix is a NxN matrix, where N is the number of cells. It represents similarity between the cells based on the averaging of clustering results from all combinations of clustering parameters. Similarity 0 (blue) means that the two cells are always assigned to different clusters. In contrast, similarity 1 (red) means that the two cells are always assigned to the same cluster. The consensus matrix is clustered by hierarchical clustering and has a diagonal-block structure. Intuitively, the perfect clustering is achieved when all diagonal blocks are completely red and all off-diagonal elements are completely blue.

**Usage**

```r
sc3_plot_consensus.SCESet(object, k, show_pdata = NULL)
```

## S4 method for signature 'SCESet'

```r
sc3_plot_consensus(object, k, show_pdata = NULL)
```
sc3_plot_expression

Arguments

object an object of 'SCESet' class
k number of clusters
p.val significance threshold used for the DE genes
show_pdata a vector of colnames of the pData(object) table. Default is NULL. If not NULL will add pData annotations to the columns of the output matrix

Description

The expression panel represents the original input expression matrix (cells in columns and genes in rows) after the gene filter. Genes are clustered by kmeans with k = 100 (dendrogram on the left) and the heatmap represents the expression levels of the gene cluster centers after log2-scaling.

Usage

sc3_plot_expression.SCESet(object, k, show_pdata = NULL)

## S4 method for signature 'SCESet'
sc3_plot_expression(object, k, show_pdata = NULL)

Arguments

object an object of 'SCESet' class
k number of clusters
show_pdata a vector of colnames of the pData(object) table. Default is NULL. If not NULL will add pData annotations to the columns of the output matrix

sc3_plot_markers

Plot expression of marker genes identified by SC3 as a heatmap.

Description

By default the genes with the area under the ROC curve (AUROC) > 0.85 and with the p-value < 0.01 are selected and the top 10 marker genes of each cluster are visualized in this heatmap.

Usage

sc3_plot_markers.SCESet(object, k, auroc = 0.85, p.val = 0.01, show_pdata = NULL)

## S4 method for signature 'SCESet'
sc3_plot_markers(object, k, auroc = 0.85, p.val = 0.01, show_pdata = NULL)
Arguments

- **object**: an object of 'SCESet' class
- **k**: number of clusters
- **auroc**: area under the ROC curve
- **p.val**: significance threshold used for the DE genes
- **show_pdata**: a vector of colnames of the pData(object) table. Default is NULL. If not NULL will add pData annotations to the columns of the output matrix

### sc3_plot_silhouette

*Plot silhouette indexes of the cells*

Description

A silhouette is a quantitative measure of the diagonality of the consensus matrix. An average silhouette width (shown at the bottom left of the silhouette plot) varies from 0 to 1, where 1 represents a perfectly block-diagonal consensus matrix and 0 represents a situation where there is no block-diagonal structure. The best clustering is achieved when the average silhouette width is close to 1.

Usage

```r
sc3_plot_silhouette.SCESet(object, k)
```

## S4 method for signature 'SCESet'

```r
sc3_plot_silhouette(object, k)
```

Arguments

- **object**: an object of 'SCESet' class
- **k**: number of clusters

### sc3_prepare

*Prepare the SCESet object for SC3 clustering.*

Description

This function prepares an object of SCESet class for SC3 clustering. It creates and populates the following items of the sc3 slot of the SCESet object:

- **kmeans_iter_max**: the same as the kmeans_iter_max argument.
- **kmeans_nstart**: the same as the kmeans_nstart argument.
- **n_dim**: contains numbers of the number of eigenvectors to be used in kmeans clustering.
- **rand_seed**: the same as the rand_seed argument.
- **svm_train_inds**: if SVM is used this item contains indexes of the training cells to be used for SC3 clustering and further SVM prediction.
* svm_study_inds - if SVM is used this item contains indexes of the cells to be predicted by SVM.
* n_cores - the same as the n_cores argument.
* ks - the same as the ks argument.

Please note that by default the "exprs" slot of the input scater object is used for the SC3 analysis. If the scater object has been created in a standard way then the expression values in the "exprs" slot will be automatically log-transformed. If you have overwritten the "exprs" slot manually, please make sure that the values in the "exprs" slot are log-transformed before running the SC3 analysis. SC3 assumes that the data is log-transformed by default.

**Usage**

```r
c3_prepare.SCESet(object, ks = NULL, gene_filter = TRUE, pct_dropout_min = 10, pct_dropout_max = 90, d_region_min = 0.04, d_region_max = 0.07, svm_num_cells = NULL, svm_train_inds = NULL, svm_max = 5000, n_cores = NULL, kmeans_nstart = NULL, kmeans_iter_max = 1e+09, rand_seed = 1)
```

**Arguments**

- **object** an object of SCESet class.
- **ks** a continuous range of integers - the number of clusters k used for SC3 clustering. Can also be a single integer.
- **gene_filter** a boolean variable which defines whether to perform gene filtering before SC3 clustering.
- **pct_dropout_min** if gene_filter = TRUE, then genes with percent of dropouts smaller than pct_dropout_min are filtered out before clustering.
- **pct_dropout_max** if gene_filter = TRUE, then genes with percent of dropouts larger than pct_dropout_max are filtered out before clustering.
- **d_region_min** defines the minimum number of eigenvectors used for kmeans clustering as a fraction of the total number of cells. Default is 0.04. See SC3 paper for more details.
- **d_region_max** defines the maximum number of eigenvectors used for kmeans clustering as a fraction of the total number of cells. Default is 0.07. See SC3 paper for more details.
- **svm_num_cells** number of randomly selected training cells to be used for SVM prediction. The default is NULL.
- **svm_train_inds** a numeric vector defining indeces of training cells that should be used for SVM training. The default is NULL.
- **svm_max** define the maximum number of cells below which SVM is not run.
sc3_run_svm

n_cores defines the number of cores to be used on the user’s machine.

kmeans_nstart nstart parameter passed to kmeans function. Default is 1000 for up to 2000 cells and 50 for more than 2000 cells.

kmeans_iter_max iter.max parameter passed to kmeans function. Default is 1e+09.

rand_seed sets the seed of the random number generator. SC3 is a stochastic method, so setting the rand_seed to a fixed values can be used for reproducibility purposes.

... further arguments passed to sc3_prepare.SCESet

Value

an object of ’SCESet’ class

sc3_run_svm Run the hybrid SVM approach.

Description

This method parallelize SVM prediction for each k (the number of clusters). Namely, for each k, support_vector_machines function is utilized to predict the labels of study cells. Training cells are selected using svm_train_indds item of the sc3 slot of the input SCESet object.

Usage

sc3_run_svm.SCESet(object)

## S4 method for signature ’SCESet’
sc3_run_svm(object)

Arguments

object an object of ’SCESet’ class

Details

Results are written to the sc3_k_clusters columns to the phenoData slot of the input object, where k is the number of clusters.

Value

an object of ’SCESet’ class
**Support Vector Machines**

Run support vector machines (SVM) prediction

### Description

Train an SVM classifier on a training dataset (`train`) and then classify a study dataset (`study`) using the classifier.

### Usage

```r
support_vector_machines(train, study, kern)
```

### Arguments

- `train`  
  Training dataset with colnames, corresponding to training labels
- `study`  
  Study dataset
- `kern`  
  Kernel to be used with SVM

### Value

Classification of the study dataset

---

**tmult**

Matrix left-multiplied by its transpose

### Description

Given matrix A, the procedure returns A'A.

### Usage

```r
tmult(x)
```

### Arguments

- `x`  
  Numeric matrix.
transformation  

**Distance matrix transformation**

**Description**

All distance matrices are transformed using either principal component analysis (PCA) or by calculating the eigenvectors of the graph Laplacian (Spectral). The columns of the resulting matrices are then sorted in descending order by their corresponding eigenvalues.

**Usage**

```r
transformation(dists, method)
```

**Arguments**

- `dists`: distance matrix
- `method`: transformation method: either `pca` or `laplacian`

**Value**

transformed distance matrix

---

**treutlein**  

*Single cell RNA-Seq data extracted from a publication by Treutlein et al.*

**Description**

Single cell RNA-Seq data extracted from a publication by Treutlein et al.

**Usage**

```r
treutlein
```

**Format**

An object of class `matrix` with 23271 rows and 80 columns.

**Source**


Columns represent cells, rows represent genes expression values. Colnames respresent indexes of cell clusters (known information based on the experimental protocol). There are 80 cells and 5 clusters in this dataset.
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