

Package ‘SigBridgeR’

December 24, 2025

Title Multi-algorithm Integration of Phenotypic, scRNA-seq, and Bulk Data for Cell Screening

Version 3.0.0

Description SigBridgeR is an integrative toolkit designed to identify phenotype-associated cell sub-populations by combining phenotype(e.g. survival, drug sensitivity), bulk expression and single-cell RNA-seq data. It leverages multiple algorithms to robustly link cell features with clinical or functional phenotypes. The package provides a unified pipeline for cross-modal data analysis, enabling the discovery of biologically and clinically relevant cell states in heterogeneous samples.

License GPL (>= 3)

Encoding UTF-8

URL <https://github.com/WangLabCSU/SigBridgeR>

BugReports <https://github.com/WangLabCSU/SigBridgeR/issues>

Roxygen list(markdown = TRUE)

RoxygenNote 7.3.3

Imports chk,
cli,
data.table,
DEGAS (>= 1.0.0),
dplyr,
edgeR,
glue,
IDConverter (>= 0.3.5),
LPSGL (>= 0.0.0.9000),
Matrix,
methods,
PIPET (>= 0.1.2),
processx,
purrr,
rlang,
scAB (>= 1.0.1),
Scissor (>= 2.0.0),
scPAS (>= 0.2.0),
ScPP (>= 0.0.0.9000),

Seurat ($\geq 5.0.0$),
 SeuratObject ($\geq 5.0.0$),
 SigBridgeRUtils ($\geq 0.1.0$),
 tibble

Depends R ($\geq 4.1.0$)

Remotes bioc::edgeR,
 github::Exceret/ALRA,
 github::Exceret/DEGAS,
 github::Exceret/LPSGL,
 github::Exceret/PIPET,
 github::Exceret/scAB,
 github::Exceret/Scissor,
 github::Exceret/scPAS,
 github::Exceret/ScPP,
 github::WangLabCSU/SigBridgeRUtils,
 github::ShixiangWang/IDConverter

Suggests ALRA,
 anndata,
 carrier,
 furr,
 future,
 future.mirai,
 ggforce,
 ggplot2,
 ggupset,
 ggVennDiagram,
 here,
 KernSmooth,
 knitr,
 matrixStats,
 matrixTests,
 org.Hs.eg.db,
 patchwork,
 preprocessCore,
 randomcoloR,
 reticulate,
 rmarkdown,
 sparseMatrixStats,
 testthat ($\geq 3.0.0$),
 tidyr,
 zeallot

VignetteBuilder knitr

Config/testthat/edition 3

Contents

AddMetaFeature	3
AddMisc	4
aggregate-dups	5
BulkPreProcess	6
DoDEGAS	9
DoLP_SGL	12
DoPIPET	14
DoscAB	16
DoScissor	17
DoscPAS	20
DoscPP	22
FindRobustElbow	24
getFuncOption	25
ListPyEnv	26
LoadRefData	28
MergeResult	29
QCFilter	30
QCPatternDetect	31
SCPreProcess	32
Screen	35
ScreenFractionPlot	39
ScreenUpset	41
setFuncOption	43
SetupPyEnv	44
SymbolConvert	45
Index	46

AddMetaFeature	<i>Add Gene-Level Metadata to Seurat Object (Vectorized, ...-based)</i>
----------------	---

Description

Add multiple feature-level metadata (vectors or 2D tables) to a Seurat object. Handles vectors (length-checked), matrices, data.frames, data.tables, etc. Columns/variables with duplicated names are suffixed (e.g., "type_1"). Gene alignment is auto-detected: rows or cols must match nrow(seurat_obj).

Usage

```
AddMetaFeature(seurat_obj, ..., assay = "RNA")
```

Arguments

seurat_obj	A Seurat object.
...	One or more metadata inputs: <ul style="list-style-type: none"> • Named/unnamed vectors (length = ngenes) • 2D objects (matrix, data.frame, data.table, etc.) where one dimension (rows or cols) has size = ngenes.
assay	Assay name (default: "RNA", fallback to first).

Value

Modified seurat_obj (invisibly).

AddMisc	<i>Safely Add Miscellaneous Data to Seurat Object</i>
---------	---

Description

Adds arbitrary data to the @misc slot of a Seurat object with automatic key conflict resolution. If the key already exists, automatically appends a numeric suffix to ensure unique key naming (e.g., "mykey_1", "mykey_2").

Usage

```
AddMisc(seurat_obj, ..., cover = TRUE)
```

Arguments

seurat_obj	A Seurat object to modify
...	key-value pairs to add to the @misc slot.
cover	Logical indicating whether to overwrite existing data. If (default TRUE).

Value

The modified Seurat object with added @misc data. The original object structure is preserved with no other modifications.

Key Generation Rules

1. If key doesn't exist: uses as-is
2. If key exists: appends the next available number (e.g., "key_1", "key_2")
3. If numbered keys exist (e.g., "key_2"): increments the highest number

Examples

```
## Not run:
# Basic usage
seurat_obj <- AddMisc(seurat_obj, "QC_stats" = qc_df)

# Auto-incrementing example
seurat_obj <- AddMisc(seurat_obj, markers = markers1)
seurat_obj <- AddMisc(seurat_obj, markers = markers2, cover=FALSE)
# Stores as "markers" and "markers_1"

## End(Not run)
```

 aggregate-dups

 Aggregate Rows or Columns with Duplicate Names

Description

These functions collapse duplicated row names (e.g., gene symbols) or column names (e.g., sample IDs) in matrix-like objects by aggregating values using configurable methods. They support:

Rows [AggregateDupRows](#): merges rows sharing the same row name.

Columns [AggregateDupCols](#): merges columns sharing the same column name.

Both [AggregateDups](#): convenience wrapper applying row-then-column aggregation.

Designed for expression matrices, count tables, or any numeric data where feature/sample duplication occurs. Handles `matrix`, `data.frame`, and S4 `Matrix` classes (e.g. `dgCMatrix`) robustly.

Convenience wrapper that first aggregates duplicated rows, then duplicated columns. Useful for cleaning matrices where both feature and sample duplication may occur.

Usage

```
AggregateDupRows(x, method = c("max", "sum", "mean", "median", "first"))
```

```
AggregateDupCols(x, method = c("max", "sum", "mean", "median", "first"))
```

```
AggregateDups(
  x,
  method = c("max", "sum", "mean", "median", "first"),
  row_method = NULL,
  col_method = NULL
)
```

Arguments

<code>x</code>	A numeric matrix-like object (see Details).
<code>method</code>	Character scalar. Aggregation method (see Methods below).
<code>row_method</code>	Aggregation method for rows. Defaults to <code>method</code> .
<code>col_method</code>	Aggregation method for columns. Defaults to <code>method</code> .

Value

An aggregated object of the same effective type as `x`, with unique row/column names.

Methods

Supported methods (applied column-wise for rows, row-wise for columns):

"max" Maximum value per group (default).
 "sum" Sum of values per group.
 "mean" Arithmetic mean (uses `na.rm = TRUE`).
 "median" Median value.
 "first" First occurrence in original order.

Input Types and Return Types

Input class	Output class (unless noted)
<code>matrix</code>	<code>matrix</code>
<code>data.frame</code>	<code>data.frame</code>
S4 Matrix	<code>matrix</code> (dense) — S4 attributes dropped for generality

Row/column order in output follows *first occurrence* of each unique name in `rownames(x)` / `colnames(x)`.

Examples

```
# Full deduplication in one step
mat <- matrix(1:16, nrow = 4,
              dimnames = list(c("TP53", "TP53", "BRCA1", "ACTB"),
                              c("S1", "S1", "S2", "S3")))
AggregateDups(mat, method = "sum")
#>      S1 S2 S3
#> TP53   5  7  9
#> BRCA1  3  7 11
#> ACTB   4  8 12
```

Description

This function performs comprehensive preprocessing and quality control analysis for bulk RNA-seq data, including data validation, filtering, batch effect detection, principal component analysis, and visualization.

Usage

```

BulkPreProcess(
  data,
  sample_info = NULL,
  gene_symbol_conversion = FALSE,
  check = TRUE,
  min_count_threshold = 10L,
  min_gene_expressed = 3L,
  min_total_reads = 1000000L,
  min_genes_detected = 10000L,
  min_correlation = 0.8,
  n_top_genes = 500L,
  show_plot_results = TRUE,
  ...
)

```

Arguments

<code>data</code>	Expression matrix with genes as rows and samples as columns, or a list containing <code>count_matrix</code> and <code>sample_info</code>
<code>sample_info</code>	Sample information data frame (optional), ignored if <code>data</code> is a list. A qualified <code>sample_info</code> should contain both <code>sample</code> and <code>condition</code> columns (case-sensitive), and there are no specific requirements for the data type stored in the <code>condition</code> column. <code>batch</code> column is optional, which is used for batch effect detection.
<code>gene_symbol_conversion</code>	Whether to convert Ensembles version IDs and TCGA version IDs to genes with <code>IDConverter</code> , default: <code>FALSE</code>
<code>check</code>	Whether to perform detailed quality checks, default: <code>TRUE</code>
<code>min_count_threshold</code>	Minimum count threshold for gene filtering. Only values greater than this threshold are considered to represent valid gene expression. Default: <code>10L</code>
<code>min_gene_expressed</code>	Minimum number of samples a gene must be expressed in, default: <code>3L</code>
<code>min_total_reads</code>	Minimum total reads per sample, default: <code>1e6L</code>
<code>min_genes_detected</code>	Minimum number of genes detected per sample, default: <code>10000</code>
<code>min_correlation</code>	Minimum correlation threshold between samples, default: <code>0.8</code>
<code>n_top_genes</code>	Number of top variable genes for PCA analysis, default: <code>500</code>
<code>show_plot_results</code>	Whether to generate visualization plots, default: <code>TRUE</code>
<code>...</code>	Additional arguments. Currently supports: <ul style="list-style-type: none"> <code>verbose</code>: Logical indicating whether to print progress messages. Defaults to <code>TRUE</code>. <code>seed</code>: For reproducibility, default is <code>123L</code>

Details

The function performs the following operations:

1. Data validation and format conversion
2. Basic statistics calculation (missing values, read depth, gene detection)
3. Optional detailed quality checks including:
 - Sample correlation analysis
 - Principal Component Analysis (PCA)
 - Outlier detection using Mahalanobis distance
 - Batch effect detection using ANOVA
4. Data filtering based on count thresholds and quality metrics
5. Optional gene symbol conversion
6. Visualization generation (PCA plots)

Value

Filtered count matrix

Quality Metrics

The function calculates and reports several quality metrics:

Data Integrity Number of missing values

Gene Count Total number of genes after filtering

Sample Read Depth Total reads per sample

Gene Detection Rate Number of genes detected per sample

Sample Correlation Pearson correlation between samples

PCA Variance Variance explained by first two principal components

Batch Effects Proportion of genes significantly affected by batch

Sample Information Format

The sample_info data frame should contain:

sample Character vector of unique sample identifiers

condition Character vector of experimental conditions

batch Optional character vector of batch identifiers

Filtering Criteria

Genes are retained if:

- They have counts \geq min_count_threshold in \geq min_gene_expressed samples

Samples are retained if:

- Total reads \geq min_total_reads
- Detected genes \geq min_genes_detected
- Mean correlation \geq min_correlation (if check=TRUE)

See Also

[cpm](#) for counts per million calculation, [prcomp](#) for PCA analysis, [cor](#) for correlation analysis, [SymbolConvert](#) for gene symbol conversion

DoDEGAS

Run DEGAS Analysis for Single-Cell and Bulk RNA-seq Data Integration

Description

This function performs DEGAS to integrate single-cell and bulk RNA-seq data, identifying phenotype-associated cells using a bootstrap aggregated multi-task learning approach.

Usage

```
DoDEGAS(
  select_fraction = 0.05,
  min_thresh = 0.4,
  matched_bulk,
  sc_data,
  phenotype = NULL,
  sc_data.pheno_colname = NULL,
  label_type = "DEGAS",
  phenotype_class = c("binary", "continuous", "survival"),
  tmp_dir = "tmp",
  env_params = list(),
  degas_params = list(),
  normality_test_method = c("jarque-bera", "d'agostino", "kolmogorov-smirnov"),
  ...
)
```

Arguments

<code>select_fraction</code>	The top percentage of selected cells will be considered as Positive cells, without considering how much larger the possible correlation coefficient of the observation group is compared to that of the control group. Only usedl when <code>phenotype_class</code> is "binary" or "survival". (default: 0.05)
<code>min_thresh</code>	DEGAS will calculate the possible correlation coefficients for each cell related to the phenotype. When the coefficient of the observation group is at least <code>min_thresh</code> larger than that of the control group, it can be considered related to the phenotype and will be marked as Positive. The priority of <code>min_thresh</code> is higher than that of <code>select_fraction</code> . (default: 0.4)
<code>matched_bulk</code>	Bulk RNA-seq data as matrix or data.frame (rows=genes, columns=samples)
<code>sc_data</code>	Single-cell data as Seurat object containing RNA assay

phenotype	Bulk-level phenotype data. For classification: binary matrix with one-hot encoding. For survival: matrix with two columns (time and event status). Can be NULL, matrix, data.frame, or vector.
sc_data.pheno_colname	Column name for single-cell phenotype in metadata (if available), default: NULL
label_type	Label type for DEGAS results (default: "DEGAS")
phenotype_class	Type of phenotype: "binary" (classification), "continuous", or "survival"
tmp_dir	Temporary directory for intermediate files (default: "tmp")
env_params	List of environment parameters for Python setup including: <ul style="list-style-type: none"> • env.name: environment name (default: "r-reticulate-degas") • env.type: environment type "conda", "environment", or "venv" (default: "environment") • env.method: environment setup method "system", "conda" (default: "system") • env.file: path to environment file (default: system.file("conda/DEGAS_environment.yml", package = "SigBridgeR")) • env.python_version: Python version (default: "3.9.15") • env.packages: named vector of Python packages and versions (default: c("tensorflow" = "2.4.1", "protobuf" = "3.20", "numpy" = "any")) • env.recreate: whether to recreate environment (default: FALSE) • env.use_conda_forge: whether to use conda-forge channel (conda only, default: TRUE) • env.verbose: verbose output (default: FALSE)
degas_params	List of DEGAS algorithm parameters including: <ul style="list-style-type: none"> • DEGAS.model_type: model type ("BlankClass", "ClassBlank", "ClassClass", "ClassCox", "BlankCox") • DEGAS.architecture: "Standard" (feed forward) or "DenseNet" (dense net), default: "DenseNet" • DEGAS.ff_depth: number of layers in model (≥ 1, default: 3) • DEGAS.pyloc: path to Python executable (default: NULL, automatic detection) • DEGAS.bag_depth: bootstrap aggregation depth (≥ 1, default: 5) • DEGAS.train_steps: training steps (default: 2000) • DEGAS.scbatch_sz: single-cell batch size (default: 200) • DEGAS.patbatch_sz: patient batch size (default: 50) • DEGAS.hidden_feats: hidden features (default: 50) • DEGAS.do_prc: dropout percentage (default: 0.5) • DEGAS.lambda1: regularization parameter 1 (default: 3.0) • DEGAS.lambda2: regularization parameter 2 (default: 3.0) • DEGAS.lambda3: regularization parameter 3 (default: 3.0) • DEGAS.seed: random seed (default: 2)
normality_test_method	Method for normality testing: "jarque-bera", "d'agostino", or "kolmogorov-smirnov"

... Additional arguments. Currently supports:

- `verbose`: Logical indicating whether to print progress messages. Defaults to TRUE.

Details

The function performs the following steps:

1. Validates input data and parameters
2. Sets up Python environment with required dependencies
3. Trains bootstrap aggregated DEGAS model using `runCCMTLBag`
4. Generates cell-level predictions using `predClassBag`
5. Applies statistical testing to identify phenotype-associated cells
6. Labels cells as "Positive" or "Other" based on selection criteria

Model type is automatically determined:

- `BlankClass`: only bulk phenotype specified (`scLab = NULL`)
- `ClassBlank`: only single-cell phenotype specified (`patLab = NULL`)
- `ClassClass`: both single-cell and bulk phenotypes specified
- `ClassCox`: single-cell phenotype + bulk survival data
- `BlankCox`: only bulk survival data specified

Value

A list containing:

- `scRNA_data`: Seurat object with DEGAS labels added to metadata
- `model`: The model trained using the input data, and it can be used for cell classification prediction.
- `DEGAS_prediction`: Data table with DEGAS predictions containing:
 - Predicted label probabilities for each cell
 - Cell labels ("Positive"/"Other") based on selection criteria
 - Difference scores for binary phenotypes
 - Cell identifiers

References

Johnson TS, Yu CY, Huang Z, Xu S, Wang T, Dong C, et al. Diagnostic Evidence GAUGE of Single cells (DEGAS): a flexible deep transfer learning framework for prioritizing cells in relation to disease. *Genome Med.* 2022 Feb 1;14(1):11.

See Also

Other screen_method: [DoLP_SGL\(\)](#), [DoPIPET\(\)](#), [DoScissor\(\)](#), [DoscAB\(\)](#), [DoscPAS\(\)](#), [DoscPP\(\)](#)

Other DEGAS: [DEGASEnvSet\(\)](#), [DEGASFindPy\(\)](#), [DEGASModelDetect\(\)](#), [DEGASParamSet\(\)](#)

Examples

```
## Not run:
# Binary classification example
result <- DoDEGAS(
  select_fraction = 0.05, # `select_fraction` only used in binary and survival phenotyping
  matched_bulk = bulk_matrix,
  sc_data = seurat_obj,
  phenotype = bulk_phenotype,
  phenotype_class = "binary"
)

# Survival analysis example
result <- DoDEGAS(
  select_fraction = 0.05, # `select_fraction` only used in binary and survival phenotyping
  matched_bulk = bulk_matrix,
  sc_data = seurat_obj,
  phenotype = survival_data,
  phenotype_class = "survival"
)

## End(Not run)
```

DoLP_SGL

Perform LP-SGL Screening Analysis

Description

Identifies phenotype-associated cell subpopulations using Lasso-Penalized Sparse Group Lasso (LP-SGL) with Leiden community detection. This method integrates bulk and single-cell RNA-seq data to identify cell subpopulations associated with phenotypic outcomes.

Usage

```
DoLP_SGL(
  matched_bulk,
  sc_data,
  phenotype,
  label_type = "LP_SGL",
  LPSGL_family = c("logit", "cox", "linear"),
  resolution = 0.6,
  alpha = 0.5,
  nfold = 5,
  dge_analysis = list(run = FALSE, logFC_threshold = 1, pval_threshold = 0.05),
  ...
)
```

Arguments

<code>matched_bulk</code>	Bulk expression matrix (features \times samples)
<code>sc_data</code>	Single-cell RNA-seq data (Seurat object)
<code>phenotype</code>	Binary phenotype vector for bulk samples
<code>label_type</code>	Character specifying phenotype label type (default: "LP_SGL")
<code>LP_SGL_family</code>	Type of regression model: "logit" (logistic), "cox" (Cox), or "linear" (linear regression)
<code>resolution</code>	Resolution parameter for Leiden clustering (default: 0.6)
<code>alpha</code>	Alpha parameter for SGL balancing L1 and L2 penalties (default: 0.5)
<code>nfold</code>	Number of folds for cross-validation (default: 5)
<code>dge_analysis</code>	List controlling differential expression analysis: <ul style="list-style-type: none"> • <code>run</code>: Whether to run DEG analysis (default: FALSE) • <code>logFC_threshold</code>: Log fold change threshold (default: 1) • <code>pval_threshold</code>: P-value threshold (default: 0.05)
<code>...</code>	Additional arguments passed to preprocessing functions, e.g.: <ul style="list-style-type: none"> • <code>verbose</code>: Whether to print progress messages (default: TRUE) • <code>seed</code>: Random seed for reproducibility (default: 123L)

Value

A list containing:

<code>scRNA_data</code>	Seurat object with LP-SGL results integrated
<code>sgl_fit</code>	Fitted SGL model object
<code>cvfit</code>	Cross-validation results
<code>dge_res</code>	Differential expression results if requested (NULL otherwise)

References

Li J, Zhang H, Mu B, Zuo H, Zhou K. Identifying phenotype-associated subpopulations through LP_SGL. *Briefings in Bioinformatics*. 2023 Nov 22;25(1):bbad424.

See Also

Other screen_method: [DoDEGAS\(\)](#), [DoPIPET\(\)](#), [DoScissor\(\)](#), [DoscAB\(\)](#), [DoscPAS\(\)](#), [DoscPP\(\)](#)

Examples

```
## Not run:
# Example using simulated data
set.seed(123)

# Create simulated data
bulk_data <- matrix(rnorm(1000*50), nrow=1000, ncol=50)
sc_data <- matrix(rnorm(1000*500), nrow=1000, ncol=500)
```

```

phenotype <- rep(c(0, 1), each=25)

# Run LP-SGL analysis
results <- DoLP_SGL(
  matched_bulk = bulk_data,
  sc_data = sc_data,
  phenotype = phenotype,
  LPSGL_family = "logit",
  resolution = 0.6,
  dge_analysis = list(run = TRUE, logFC_threshold = 1, pval_threshold = 0.05)
)

# Access results
lpsgl_seurat <- results$scRNA_data
sgl_model <- results$sgl_fit
dge_results <- results$dge_res

## End(Not run)

```

DoPIPET

Perform PIPET Screening Analysis

Description

Predicts cell subpopulations in single-cell data by matching expression profiles to predefined marker gene templates using various distance/similarity metrics. This function implements a template-based classification approach with permutation testing for significance assessment.

Usage

```

DoPIPET(
  matched_bulk,
  sc_data,
  phenotype,
  phenotype_class = c("binary", "continuous"),
  group = NULL,
  discretize_method = c("kmeans", "median", "custom"),
  cutoff = NULL,
  label_type = "PIPET",
  log2FC = 1L,
  p_adjust = 0.05,
  show_log2FC = TRUE,
  freq_counts = NULL,
  normalize = TRUE,
  scale = TRUE,
  nPerm = 1000L,
  distance = c("cosine", "pearson", "spearman", "kendall", "euclidean", "maximum"),
  ...
)

```

Arguments

matched_bulk	Normalized bulk expression matrix (features \times samples). Column names must match phenotype identifiers.
sc_data	Seurat object containing single-cell RNA-seq data.
phenotype	Clinical outcome data. Can be: - Vector: named with sample IDs - Data frame: with row names matching bulk columns
phenotype_class	Analysis mode: - "binary": Case-control design (e.g., responder/non-responder) - "continuous": Continuous outcome (e.g.,)
group	A character, name of one metadata column to group cells by (for example, orig.ident). The default value is NULL. In this case, screening will be performed on each group separately.
discretize_method	c("median", "kmeans", "custom"). Discretization strategy for continuous phenotypes. Note: "median" is mapped internally to "quantile" (2-group quantile split). Default: "median".
cutoff	Numeric vector of length $n_{\text{group}} - 1$. Required only when discretize_method = "custom". Defines interior breakpoints on the <i>normalized, log2-transformed scale</i> (i.e., after $\text{scale}(\log_2(x + 1))$). Must be sorted in ascending order.
label_type	Character specifying phenotype label type (e.g., "SBS1", "time"), stored in scRNA_data@misc
log2FC	In the DESeq differential expression analysis results, the cutoff value of log2FC. The default value is 1L.
p_adjust	In the DESeq differential expression analysis results, the cutoff value of adjust P. The default value is 0.05.
show_log2FC	Select whether to show log2 fold changes. The default value is TRUE.
freq_counts	An integer, keep genes expressed in more than a certain number of cells. The default value is NULL, which means no filtering.
normalize	Select whether to perform normalization of count data. The default value is TRUE.
scale	Select whether to scale and center features in the dataset. The default value is TRUE.
nPerm	An integer, number of permutations to do. The default value is 1000L.
distance	A character, the distance algorithm must be included in "cosine", "pearson", "spearman", "kendall", "euclidean", "maximum". default value is NULL, which means "cosine".
...	Additional arguments to be passed to PIPET.optimized. <ul style="list-style-type: none"> • seed: Random seed for reproducibility • verbose: Whether to show progress messages • parallel: Whether to use parallel processing, default is FALSE. future::plan() must be set before calling this function.

See Also

Other screen_method: [DoDEGAS\(\)](#), [DoLP_SGL\(\)](#), [DoScissor\(\)](#), [DoscAB\(\)](#), [DoscPAS\(\)](#), [DoscPP\(\)](#)

DoscAB

*Perform scAB Screening Analysis***Description**

Implements the scAB algorithm to identify phenotype-associated cell subpopulations in single-cell RNA-seq data by integrating matched bulk expression and phenotype information. Uses non-negative matrix factorization (NMF) with dual regularization for phenotype association and cell-cell similarity.

Usage

```
DoscAB(
  matched_bulk,
  sc_data,
  phenotype,
  label_type = "scAB",
  phenotype_class = c("binary", "survival"),
  alpha = c(0.005, NULL),
  alpha_2 = c(0.005, NULL),
  maxiter = 2000L,
  tred = 2L,
  ...
)
```

Arguments

matched_bulk	Normalized bulk expression matrix (genes \times samples) where: - Columns match phenotype row names - Genes match features in sc_data
sc_data	Seurat object containing preprocessed single-cell data:
phenotype	Data frame with clinical annotations where: - Rows correspond to matched_bulk columns - For survival: contains time and status columns
label_type	Character specifying phenotype label type (e.g., "SBS1", "time"), stored in scRNA_data@misc
phenotype_class	Analysis mode: - "binary": Case-control design (e.g., responder/non-responder) - "survival": Time-to-event analysis data.frame
alpha	Coefficient of phenotype regularization (default=0.005).
alpha_2	Coefficient of cell-cell similarity regularization (default=0.005).
maxiter	Maximum number of iterations for NMF (default=2000).
tred	Z-score threshold in finding subsets (default=2).
...	Additional arguments. Currently supports: <ul style="list-style-type: none"> • verbose: Logical indicating whether to print progress messages. Defaults to TRUE. • seed: For reproducibility, default is 123L

Value

A list containing:

scRNA_data Filtered Seurat object with selected cells

scAB_result scAB screening result

LICENSE

Licensed under the GNU General Public License version 3 (GPL-3.0). A copy of the license is available at <https://www.gnu.org/licenses/gpl-3.0.en.html>.

References

Zhang Q, Jin S, Zou X. scAB detects multiresolution cell states with clinical significance by integrating single-cell genomics and bulk sequencing data. *Nucleic Acids Research*. 2022 Nov 28;50(21):12112–30.

See Also

Other screen_method: [DoDEGAS\(\)](#), [DoLP_SGL\(\)](#), [DoPIPET\(\)](#), [DoScissor\(\)](#), [DoscPAS\(\)](#), [DoscPP\(\)](#)

Examples

```
## Not run:
# Binary phenotype example
result <- DoscAB(
  matched_bulk = bulk_matrix,
  sc_data = seurat_obj,
  phenotype = clinical_df,
  label_type = "disease_status",
  phenotype_class = "binary",
  alpha = 0.005,
  alpha_2 = 0.005,
  maxiter = 2000,
  tred = 2
)

## End(Not run)
```

Description

Identifies phenotype-associated cell subpopulations in single-cell data using regularized regression on matched bulk expression profiles. Scissor integrates bulk and single-cell RNA-seq data to identify cells that are significantly associated with phenotypic outcomes.

Usage

```

DoScissor(
  path2load_scissor_cache = NULL,
  path2save_scissor_inputs = "Scissor_inputs.RData",
  matched_bulk,
  sc_data,
  phenotype,
  label_type = "scissor",
  alpha = c(0.05, NULL),
  cutoff = 0.2,
  scissor_family = c("gaussian", "binomial", "cox"),
  reliability_test = list(
    run = FALSE, # whether to run reliability test
    n = 10L, # permutation times
    nfold = 10L # cross validation folds
  ),
  cell_evaluation = list(
    run = FALSE, # whether to run cell evaluation
    benchmark_data = "path_to_file.RData", # path to benchmark data
    FDR_cutoff = 0.05,
    bootstrap_n = 100L
  ),
  ...
)

```

Arguments

<code>path2load_scissor_cache</code>	Path to precomputed Scissor inputs (RData file). If provided, skips recomputation (default: NULL).
<code>path2save_scissor_inputs</code>	Path to save intermediate files (default: "Scissor_inputs.RData").
<code>matched_bulk</code>	Normalized bulk expression matrix (features \times samples). Column names must match phenotype identifiers.
<code>sc_data</code>	Seurat object containing single-cell RNA-seq data.
<code>phenotype</code>	Clinical outcome data. Can be: - Vector: named with sample IDs - Data frame: with row names matching bulk columns
<code>label_type</code>	Character specifying phenotype label type (e.g., "SBS1", "time"), stored in <code>scRNA_data@misc</code>
<code>alpha</code>	Parameter used to balance the effect of the l1 norm and the network-based penalties. It can be a number or a searching vector. If <code>alpha = NULL</code> , a default searching vector is used. The range of alpha is between 0 and 1. A larger alpha lays more emphasis on the l1 norm.
<code>cutoff</code>	(default: 0.2). When <code>alpha=NULL</code> , the cutoff is used to determine the optimal alpha. Higher values increase specificity.
<code>scissor_family</code>	Model family for outcome type: - "gaussian": Continuous outcomes - "binomial": Binary outcomes (default) - "cox": Survival outcomes

```

reliability_test
    List controlling reliability testing:
    • run: Whether to perform reliability test (default: FALSE)
    • n: Permutation times (default: 10L)
    • nfold: Cross-validation folds (default: 10L)
cell_evaluation
    List controlling cell evaluation:
    • run: Whether to perform cell evaluation (default: FALSE)
    • benchmark_data: Path to benchmark data (RData file)
    • FDR_cutoff: FDR threshold for evaluation (default: 0.05)
    • bootstrap_n: Bootstrap iterations (default: 100L)
...
    Additional arguments. Currently supports:
    • verbose: Logical indicating whether to print progress messages. Defaults
      to TRUE.
    • seed: For reproducibility, default is 123L

```

Value

A list containing:

scRNA_data A Seurat object with screened cells containing metadata:

scissor "Positive"/"Negative"/"Neutral" classification

label_type Outcome label used

scissor_result Raw Scissor results

reliability_result If reliability_test=TRUE, contains:

statistic A value between 0 and 1

p p-value of the test statistic

AUC_test_real 10 values of AUC for real data

AUC_test_back A list of AUC for background data

cell_evaluation If cell_evaluation=TRUE, contains:

evaluation_res A data.frame with some supporting information for each Scissor selected cell

LICENSE

Licensed under the GNU General Public License version 3 (GPL-3.0). A copy of the license is available at <https://www.gnu.org/licenses/gpl-3.0.en.html>.

References

Sun D, Guan X, Moran AE, Wu LY, Qian DZ, Schedin P, et al. Identifying phenotype-associated subpopulations by integrating bulk and single-cell sequencing data. Nat Biotechnol. 2022 Apr;40(4):527–38.

See Also

Other screen_method: [DoDEGAS\(\)](#), [DoLP_SGL\(\)](#), [DoPIPET\(\)](#), [DoscAB\(\)](#), [DoscPAS\(\)](#), [DoscPP\(\)](#)

Other scissor: [DoScissorCellEval\(\)](#), [DoScissorRelTest\(\)](#)

Examples

```
## Not run:
# Binary outcome example
res <- DoScissor(
  matched_bulk = bulk_matrix,
  sc_data = seurat_obj,
  phenotype = a_named_vector,
  scissor_family = "binomial"
)

## End(Not run)
```

DoscPAS

Perform scPAS Screening Analysis

Description

This function performs scPAS screening analysis by integrating bulk and single-cell RNA-seq data. It includes data filtering steps and wraps the core `scPAS::scPAS` function.

Usage

```
DoscPAS(
  matched_bulk,
  sc_data,
  phenotype,
  label_type = "scPAS",
  assay = "RNA",
  imputation = FALSE,
  imputation_method = c("KNN", "ALRA"),
  nfeature = 3000L,
  alpha = c(0.01, NULL),
  cutoff = 0.2,
  network_class = c("SC", "bulk"),
  scPAS_family = c("cox", "gaussian", "binomial"),
  permutation_times = 2000L,
  FDR_threshold = 0.05,
  independent = TRUE,
  ...
)
```

Arguments

<code>matched_bulk</code>	Bulk RNA-seq data (genes x samples)
<code>sc_data</code>	Single-cell RNA-seq data (Seurat object and preprocessed)
<code>phenotype</code>	Phenotype data frame with sample annotations

label_type	Character specifying phenotype label type (e.g., "SBS1", "time"), stored in <code>scRNA_data@misc</code> . (default: "scPAS")
assay	Assay to use from <code>sc_data</code> (default: 'RNA')
imputation	Logical, whether to perform imputation (default: FALSE)
imputation_method	Character. Name of alternative method for imputation. (options: "KNN", "ALRA")
nfeature	Number of features to select (default: 3000, indicating that the top 3000 highly variable genes are selected for model training)
alpha	Numeric. Significance threshold. Parameter used to balance the effect of the l1 norm and the network-based penalties. It can be a number or a searching vector. If <code>alpha = NULL</code> , a default searching vector is used. The range of <code>alpha</code> is in $[0, 1]$. A larger <code>alpha</code> lays more emphasis on the l1 norm. (default: 0.01)
cutoff	Numeric. Cutoff value for selecting the optimal <code>alpha</code> value when <code>alpha = NULL</code> . (default: 0.2)
network_class	Network class to use (default: 'SC', indicating gene-gene similarity networks derived from single-cell data. The other one is 'bulk'.)
scPAS_family	Model family for analysis (options: "cox", "gaussian", "binomial")
permutation_times	Number of permutations to perform (default: 2000)
FDR_threshold	Numeric. FDR value threshold for identifying phenotype-associated cells (default: 0.05)
independent	Logical. The background distribution of risk scores is constructed independently of each cell. (default: TRUE)
...	Additional arguments. Currently supports: <ul style="list-style-type: none"> • <code>verbose</code>: Logical indicating whether to print progress messages. Defaults to TRUE. • <code>seed</code>: For reproducibility, default is 123L

Value

A Seurat object from scPAS analysis

LICENSE

Licensed under the GNU General Public License version 3 (GPL-3.0). A copy of the license is available at <https://www.gnu.org/licenses/gpl-3.0.en.html>.

References

Xie A, Wang H, Zhao J, Wang Z, Xu J, Xu Y. scPAS: single-cell phenotype-associated subpopulation identifier. *Briefings in Bioinformatics*. 2024 Nov 22;26(1):bbae655.

See Also

Other screen_method: `DoDEGAS()`, `DoLP_SGL()`, `DoPIPET()`, `DoScissor()`, `DoscAB()`, `DoscPP()`

DoscPP

*Perform scPP screening analysis***Description**

This function performs scPP screening on single-cell data using matched bulk data and phenotype information. It supports binary, continuous, and survival phenotype types.

Usage

```
DoscPP(
  matched_bulk,
  sc_data,
  phenotype,
  label_type = "scPP",
  phenotype_class = c("Binary", "Continuous", "Survival"),
  ref_group = 0,
  Log2FC_cutoff = 0.585,
  estimate_cutoff = 0.2,
  probs = c(0.2, NULL),
  ...
)
```

Arguments

matched_bulk	Bulk expression data (genes × samples) where: - Column names must match phenotype row names
sc_data	Seurat object containing preprocessed single-cell data: - Normalized counts in RNA assay
phenotype	Data frame or tibble or named vector with: - Rownames matching matched_bulk columns - For survival: must contain time and status columns
label_type	Character specifying phenotype label type (e.g., "SBS1"), stored in scRNA_data@misc
phenotype_class	Analysis type (case-sensitive): - "Binary": Case-control studies (e.g., tumor/normal) - "Continuous": Quantitative traits (e.g., drug response) - "Survival": Time-to-event data (requires time/status columns)
ref_group	Reference group or baseline for binary comparisons, e.g. "Normal" for Tumor/Normal studies and 0 for 0/1 case-control studies. (default: 0)
Log2FC_cutoff	Minimum log2 fold-change for binary markers (default: 0.585)
estimate_cutoff	Effect size threshold for continuous traits (default: 0.2)
probs	A numeric value indicating the quantile cutoff for cell classification. This parameter can also be a numeric vector, in which case an optimal threshold will be selected based on the AUC and enrichment score.(default: 0.2)

... Additional arguments. Currently supports:

- **verbose**: Logical indicating whether to print progress messages. Defaults to TRUE.
- **seed**: For reproducibility, default is 123L

Value

A list containing:

scRNA_data Seurat object with added metadata:

ScPP "Positive"/"Negative"/"Neutral" classification

gene_list List of genes used for screening

Algorithm Steps

1. Data Validation: Checks sample alignment between bulk and phenotype data
2. Marker Selection: Identifies phenotype-associated genes from bulk data
3. Single-cell Screening: Projects bulk markers onto single-cell data
4. Cell Classification: Categorizes cells based on phenotype association

Reference

WangX-Lab/ScPP [Internet]. [cited 2025 Aug 31]. Available from: <https://github.com/WangX-Lab/ScPP>

See Also

Other screen_method: [DoDEGAS\(\)](#), [DoLP_SGL\(\)](#), [DoPIPET\(\)](#), [DoScissor\(\)](#), [DoscAB\(\)](#), [DoscPAS\(\)](#)

Examples

```
## Not run:
# Binary phenotype analysis
res <- DoscPP(
  matched_bulk = bulk_data,
  sc_data = seurat_obj,
  phenotype = ms_data,
  label_type = "SBS1",
  phenotype_class = "Binary"
)

# Survival analysis
surv_res <- DoscPP(
  sc_data = seurat_obj,
  matched_bulk = bulk_data,
  phenotype = surv_df,
  label_type = "OS_status",
  phenotype_class = "Survival"
)
```

```
## End(Not run)
```

FindRobustElbow	<i>Automatically determine optimal PCA dimensions using multiple robust methods</i>
-----------------	---

Description

This function combines multiple statistical approaches to automatically determine the optimal number of principal components (PCs) for downstream single-cell analysis. It integrates variance-based heuristics, elbow detection algorithms, and provides comprehensive visualization for result validation.

Usage

```
FindRobustElbow(
  obj,
  verbose = SigBridgeUtils::getFuncOption("verbose"),
  ndims = 50L
)
```

Arguments

obj	A Seurat object that has PCA computed (after RunPCA)
verbose	Logical, if TRUE outputs detailed method results and creates visualization plot. If FALSE returns only the final dimension.
ndims	Integer, maximum number of dimensions to consider (default: 50L)

Value

Integer, the recommended number of PCA dimensions for downstream analysis

See Also

Other single_cell_preprocess: [FilterTumorCell\(\)](#), [Pattern2Colname\(\)](#), [ProcessSeuratObject\(\)](#), [QCPatternDetect\(\)](#)

Examples

```
## Not run:
# After running PCA on Seurat object
pbmc <- RunPCA(pbmc, npcs = 50)
optimal_dims <- FindRobustElbow(pbmc, verbose = TRUE)
pbmc <- FindNeighbors(pbmc, dims = 1:optimal_dims)

## End(Not run)
```


Description

These functions provide a centralized configuration system for the SigBridgeR package, allowing users to set and retrieve package-specific options with automatic naming conventions.

setFuncOption sets one or more configuration options for the SigBridgeR package. Options are automatically prefixed with "SigBridgeR." if not already present, ensuring proper namespace isolation.

getFuncOption retrieves configuration options for the SigBridgeR package. The function automatically handles the "SigBridgeR." prefix, allowing users to reference options with or without the explicit prefix.

checkFuncOption validates configuration options for the SigBridgeR package. This function ensures that all options meet type and value requirements before they are set in the global options.

Usage

```
getFuncOption(option, default = NULL)
```

Arguments

option	Character string specifying the option name to check
default	The default value to return if the option is not set. Defaults to NULL.

Details

This function performs the following validations:

verbose, parallel Must be single logical values (TRUE/FALSE)

parallel.type Must be a single character string

workers, timeout, seed Must be single integer values

The function is automatically called by setFuncOption to ensure all configuration options are valid before they are set.

Value

Invisibly returns NULL. The function is called for its side effects of setting package options.

The value of the specified option if it exists, otherwise the provided default value.

No return value. The function throws an error if the value doesn't meet the required specifications for the given option.

Available Options

- `verbose`: A logical value indicating whether to print verbose messages, defaults to `TRUE`
- `timeout`: An integer specifying the timeout in seconds for parallel processing, defaults to `'180L'`
- `seed`: An integer specifying the random seed for reproducible results, defaults to `'123L'`

Examples

```
# Set options with automatic prefixing
setFuncOption(verbose = TRUE)

# Retrieve options with automatic prefixing
getFuncOption("verbose")
```

ListPyEnv

List Available Python Environments

Description

Discovers and lists available Python environments of various types on the system. This generic function provides a unified interface to find Conda environments and virtual environments (venv) through S3 method dispatch.

Default method that lists all Python environments by combining results from Conda and virtual environment discovery methods.

Discovers Conda environments using multiple detection strategies for maximum reliability. First attempts to use system Conda commands, then falls back to reticulate's built-in Conda interface if Conda command is unavailable or fails. Returns empty data frame if Conda is not available or no environments are found.

Discovers virtual environments by searching common venv locations including user directories (`'~/virtualenvs'`, `'~/venvs'`) and project folders (`'./venv'`, `'./venv'`). Supports custom search paths through the `'venv_locations'` parameter. Returns empty data frame if no virtual environments are found in the specified locations.

Usage

```
ListPyEnv(
  env_type = c("all", "conda", "venv", "virtualenv"),
  timeout = 30000,
  venv_locations = c("~/virtualenvs", "~/venvs", "./venv", "./venv"),
  verbose = TRUE,
  ...
)
```

Arguments

<code>env_type</code>	Character string specifying the type of environments to list. One of: <code>"all"</code> , <code>"conda"</code> , <code>"venv"</code> . Defaults to <code>"all"</code> .
<code>timeout</code>	The maximum timeout time when using system commands, only effective when <code>'env_type=conda'</code> .
<code>venv_locations</code>	Character vector of directory paths to search for virtual environments. Default includes standard locations and common project directories.
<code>verbose</code>	Logical indicating whether to print verbose output.
<code>...</code>	For future use.

Details

The function uses S3 method dispatch to handle different environment types:

- `***"all"***`: Combines results from all environment types using `'rbind()'` - `***"conda"***`: Searches for Conda environments using multiple methods: - Primary: `'reticulate::conda_list()'` for reliable environment detection - Fallback: System `'conda info --envs'` command for broader compatibility - `***"venv"***`: Searches common virtual environment locations including user directories and project folders

Each method includes comprehensive error handling and will return empty results with informative warnings if no environments are found or if errors occur during discovery.

Value

A data frame with the following columns:

- `'name'` - Character vector of environment names
- `'python'` - Character vector of paths to Python executables
- `'type'` - Character vector indicating environment type (`"conda"` or `"venv"`)

Returns an empty data frame with these columns if no environments are found.

Examples

```
## Not run:
# List all Python environments
ListPyEnv("all")

# List only Conda environments
ListPyEnv("conda")

# List only virtual environments with custom search paths
ListPyEnv("venv", venv_locations = c("~/my_envs", "./project_env"))

## End(Not run)
```

LoadRefData*Download & Load Reference Data*

Description

This function checks if the data already exists in a local cache. If not, it downloads the file from the remote repository using multiple sources with fallback.

Usage

```
LoadRefData(  
  data_type = c("survival", "binary", "continuous"),  
  path = NULL,  
  timeout = SigBridgeUtils::getFuncOption("timeout"),  
  ...  
)
```

Arguments

<code>data_type</code>	The type of data to download. Must be one of "survival", "binary", or "continuous".
<code>path</code>	Optional path to save the downloaded file, default: NULL, saving in package.
<code>timeout</code>	Integer. Connection timeout in seconds (default: 60)
<code>...</code>	Additional arguments. Currently supports: <ul style="list-style-type: none">• <code>verbose</code>: Logical indicating whether to print progress messages. Defaults to TRUE.

Value

The requested datasets, stored in a list.

Examples

```
## Not run:  
ref_data <- LoadRefData(data_type = c("survival"))  
  
## End(Not run)
```

MergeResult

*Merge Multiple Screening Analysis Results***Description**

Combines results from multiple single-cell screening analyses (Scissor, scPAS, scPP, or scAB) by merging their metadata and miscellaneous information while preserving the original expression data. Performs an inner join on cell barcodes to ensure only cells present in all inputs are retained.

Usage

```
MergeResult(..., verbose = SigBridgeUtils::getFuncOption("verbose"))
```

Arguments

...	Input objects to merge. Can be: - Seurat objects - Lists containing scRNA_data (Seurat objects) - Mixed combinations of the above - The first one will be used as base object for merging, priority given to first one when duplicate columns are found
verbose	Logical, whether to print a message

Value

A merged Seurat object containing:

- Expression data from the first input object
- Combined metadata from all input objects
- Miscellaneous information from all input objects
- Only cells present in all input objects (inner join)

Processing Details

1. Input Validation: Checks for valid Seurat objects or lists containing Seurat objects
2. Metadata Extraction: Collects metadata from all objects
3. Cell Intersection: Retains only cells present in all datasets
4. Object Merging: Creates new Seurat object with combined metadata
5. Miscellaneous: Adds miscellaneous information to the merged object

Examples

```
## Not run:
# Merge mixed analysis types
combined <- MergeResult(scissor_output, scAB_output, scPP_output)

# Merge list-containing objects
merged_list <- MergeResult(list1, list2, seurat_obj)
```

```
## End(Not run)
```

QCFilter

Filter Seurat object cells by QC metrics

Description

Filters cells based on nFeature_RNA and optional QC metrics (e.g. percent.mt, percent.rp), defined in `seurat_obj@misc$qc_colnames` (See [QCPatternDetect](#)). Only metrics with non-constant, non-all-zero values are used.

Usage

```
QCFilter(
  seurat_obj,
  data_filter.thresh = list(nFeature_RNA_thresh = c(200L, 6000L), nCount_RNA_thresh =
    c(500L, 50000L), percent.mt = 20L, percent.rp = 60L),
  verbose = TRUE,
  ...
)
```

Arguments

<code>seurat_obj</code>	A Seurat object.
<code>data_filter.thresh</code>	A named list with thresholds. Default: <code>list(nFeature_RNA_thresh = c(200L, 6000L), nCount_RNA_thresh = c(500L, 25000L), percent.mt = 20L, percent.rp = 60L)</code> .
<code>verbose</code>	Logical; whether to print progress messages via cli.
<code>...</code>	No use

Value

A filtered Seurat object.

QCPatternDetect

Calculate Percentage of Features Matching Patterns

Description

This function calculates the percentage of counts coming from features matching specified patterns (e.g., mitochondrial genes, ribosomal genes) and adds them as metadata columns to the Seurat object.

Usage

```
QCPatternDetect(
  obj,
  pattern = c("^MT-", "^mt-", "^RP[SL]", "^MT-|^RP[SL]"),
  verbose = TRUE,
  ...
)
```

Arguments

obj	A seurat object.
pattern	A character vector or list containing regex patterns to identify mitochondrial genes, ribosomal protein genes, or other unwanted genes, as well as combinations of these genes. Customized patterns are supported.
verbose	logical, whether to print progress messages
...	Additional arguments passed to PercentageFeatureSet

Details

The function automatically generates friendly column names based on the patterns:

- "mt" for mitochondrial patterns
- "rp" for ribosomal patterns
- "rrna" for ribosomal RNA patterns
- For combined patterns (using |), creates names like "mt_rp"
- For other patterns, creates cleaned lowercase names

See Also

Other single_cell_preprocess: [FilterTumorCell\(\)](#), [FindRobustElbow\(\)](#), [Pattern2Colname\(\)](#), [ProcessSeuratObject\(\)](#)

SCPreProcess

Single-Cell RNA-seq Preprocessing Pipeline

Description

A generic function for standardized preprocessing of single-cell RNA-seq data from multiple sources. Handles data.frame/matrix, AnnData, and Seurat inputs with tumor cell filtering. Implements a complete analysis pipeline from raw data to clustered embeddings.

Usage

```
SCPreProcess(sc, ...)
```

```
## Default S3 method:
```

```
SCPreProcess(
  sc,
  meta_data = NULL,
  column2only_tumor = NULL,
  project = "SC_Screening_Proj",
  min_cells = 400L,
  min_features = 0L,
  quality_control = list(pattern = c("^MT-")),
  data_filter = list(nFeature_RNA_thresh = c(200L, 6000L), nCount_RNA_thresh = c(500L,
    50000L), percent.mt = 20L, percent.rp = 60L),
  normalization_method = "LogNormalize",
  scale_factor = 10000L,
  scale_features = NULL,
  selection_method = "vst",
  resolution = 0.6,
  dims = NULL,
  ...
)
```

```
## S3 method for class 'R6'
```

```
SCPreProcess(
  sc,
  meta_data = NULL,
  column2only_tumor = NULL,
  project = "SC_Screening_Proj",
  min_cells = 400L,
  min_features = 0L,
  quality_control = list(pattern = c("^MT-")),
  data_filter = list(nFeature_RNA_thresh = c(200L, 6000L), nCount_RNA_thresh = c(500L,
    50000L), percent.mt = 20L, percent.rp = 60L),
  normalization_method = "LogNormalize",
  scale_factor = 10000L,
  scale_features = NULL,
```



```

    selection_method = "vst",
    resolution = 0.6,
    dims = NULL,
    ...
)

## S3 method for class 'Seurat'
SCPreProcess(sc, column2only_tumor = NULL, ...)

```

Arguments

<code>sc</code>	Input data, one of: <ul style="list-style-type: none"> • <code>data.frame/matrix/dgCMatrix</code>: Raw count matrix (features x cells) • R6: Python <code>AnnData</code> object, obtained via package <code>anndata</code> or <code>anndataR</code> • <code>Seurat</code>: Preprocessed <code>Seurat</code> object
<code>...</code>	Additional arguments passed to specific methods. Currently supports: <ul style="list-style-type: none"> • <code>verbose</code>: Logical indicating whether to print progress messages. Defaults to <code>TRUE</code>. • <code>dims_Neighbors</code>: Dimensions to use for <code>FindNeighbors</code>. Defaults to <code>NULL</code>, using <code>dims</code>. • <code>dims_TSNE</code>: Dimensions to use for <code>RunTSNE</code>. Defaults to <code>NULL</code>, using <code>dims</code>. • <code>dims_UMAP</code>: Dimensions to use for <code>RunUMAP</code>. Defaults to <code>NULL</code>, using <code>dims</code>.
<code>meta_data</code>	A <code>data.frame</code> containing metadata for each cell. It will be added to the <code>Seurat</code> object as <code>@meta.data</code> . If <code>sc</code> is an <code>anndata</code> object, <code>obs</code> will be automatically used.
<code>column2only_tumor</code>	A character of column names in <code>meta_data</code> , used to filter the <code>Seurat</code> object to only tumor cells. If <code>NULL</code> , no filtering is performed.
<code>project</code>	A character of project name, used to name the <code>Seurat</code> object.
<code>min_cells</code>	Minimum number of cells that must express a feature for it to be included in the analysis. Defaults to 400L.
<code>min_features</code>	Minimum number of features that must be detected in a cell for it to be included in the analysis. Defaults to 0L.
<code>quality_control</code>	A list of QC settings. If <code>NULL</code> , no QC metrics are computed. Default: <code>list(pattern = "^MT-")</code> . pattern A character vector of regex patterns to identify gene groups (e.g., mitochondrial, ribosomal).
<code>data_filter</code>	A list of filtering thresholds. If <code>NULL</code> , no cell filtering is performed. Default: <code>list(nFeature_RNA_thresh = c(200L, 6000L), nCount_RNA_thresh = c(500L, 50000L), percent_mt = 20L, percent_rp = 60L)</code>
<code>normalization_method</code>	Method for normalization: "LogNormalize", "CLR", or "RC". Defaults to "LogNormalize".
<code>scale_factor</code>	Scaling factor for normalization. Defaults to 10000L.
<code>scale_features</code>	Features to use for scaling. If <code>NULL</code> , uses all variable features. If "hvg", uses high-variance genes via <code>VariableFeatures()</code> . Defaults to <code>NULL</code> .

selection_method	Method for variable feature selection: "vst", "mvp", or "disp". Defaults to "vst".
resolution	Resolution parameter for clustering. Higher values lead to more clusters. Defaults to 0.6.
dims	Dimensions to use for clustering and dimensionality reduction. If NULL, automatically determined by elbow method. Defaults to NULL.

Details

Quality Control Patterns: The function supports flexible pattern matching for quality control, for example:

- `"^MT-"` - Mitochondrial genes (default)
- `"^RP\[LS\]"` - Ribosomal protein genes
- `"^\[rt\]rna"` - rRNA and tRNA genes
- Custom patterns using regular expressions
- Combined patterns: `"^MT-|^RP\[LS\]"` for both mitochondrial and ribosomal genes

Flexible Filtering: The filtering system dynamically adapts to detected quality control patterns:

- Column names are automatically generated from patterns
- Multiple thresholds can be specified in `data_filter.thresh`
- Use `SigBridgeR::Pattern2Colname` to determine correct column names for custom patterns if still confused

Value

A Seurat object containing:

- Normalized and scaled expression data
- Variable features identified by selection method
- PCA, t-SNE, and UMAP dimensionality reductions
- Cluster identities at specified resolution
- Quality control metrics in `@meta.data`
- When tumor cells filtered: original dimensions in `@misc$raw_dim`
- Final dimensions in `@misc$self_dim`
- Quality control column names in `@misc$qc_colnames`

See Also

[CreateSeuratObject](#), [NormalizeData](#), [ScaleData](#), [FindVariableFeatures](#), [RunPCA](#), [RunTSNE](#), [RunUMAP](#), [FindNeighbors](#), [FindClusters](#)

Examples

```
## Not run:
# Example with matrix input
counts_matrix <- matrix(rpois(1000, 5), nrow = 100, ncol = 10)
rownames(counts_matrix) <- paste0("Gene", 1:100)
colnames(counts_matrix) <- paste0("Cell", 1:10)

seurat_obj <- SCPreProcess(
  sc = counts_matrix,
  project = "TestProject",
  min_features = 50,
  resolution = 0.8
)

# Example with tumor cell filtering
metadata <- data.frame(
  cell_type = c(rep("Tumor", 5), rep("Normal", 5)),
  row.names = paste0("Cell", 1:10)
)

tumor_seurat <- SCPreProcess(
  sc = counts_matrix,
  meta_data = metadata,
  column2only_tumor = "cell_type",
  project = "TumorAnalysis"
)

## End(Not run)
```

Screen

Single-Cell Data Screening

Description

Integrates matched bulk expression data and phenotype information to identify phenotype-associated cell populations in single-cell RNA-seq data using one of four computational methods. Ensures consistency between bulk and phenotype data before analysis.

Usage

```
Screen(
  matched_bulk,
  sc_data,
  phenotype,
  label_type = NULL,
  phenotype_class = c("binary", "survival", "continuous"),
  screen_method = c("Scissor", "scPP", "scPAS", "scAB", "DEGAS", "LP_SGL", "PIPET"),
  ...
)
```

Arguments

<code>matched_bulk</code>	Matrix or data frame of preprocessed bulk RNA-seq expression data (genes x samples). Column names must match names/IDs in phenotype.
<code>sc_data</code>	A Seurat object containing scRNA-seq data to be screened.
<code>phenotype</code>	Phenotype data, either: - Named vector (names match <code>matched_bulk</code> columns), or - Patient survival Data frame with row names matching <code>matched_bulk</code> columns, colnames named "time" and "status"
<code>label_type</code>	Character specifying phenotype label type (e.g., "SBS1", "time")
<code>phenotype_class</code>	Type of phenotypic outcome (must be consistent with input data): - "binary": Binary traits (e.g., case/control) - "continuous": Continuous measurements - "survival": Survival information
<code>screen_method</code>	Screening algorithm to use, there are six options: <ul style="list-style-type: none"> • "Scissor": see also <code>DoScissor()</code> • "scPP": see also <code>DoscPP()</code> • "scPAS": see also <code>DoscPAS()</code> • "scAB": see also <code>DoscAB()</code>, continuous phenotype is not supported • "DEGAS": see also <code>DoDEGAS()</code> • "LP_SGL": see also <code>DoLP_SGL()</code>
<code>...</code>	Additional method-specific parameters: <p>Scissor alpha (numeric or NULL) Significance threshold. When NULL, alpha will keep increasing iteratively until the corresponding cells are screened out, default 0.05</p> <p>cutoff (numeric) A threshold for terminating the iteration of alpha, only work when alpha is NULL, default 0.2</p> <p>path2load_scissor_cache (character) default NULL</p> <p>path2save_scissor_inputs (character) A path to save the intermediary data. By using <code>path2load_scissor_cache</code>, the intermediary data can be loaded from the specified path. default "Scissor_inputs.RData"</p> <p>reliability_test (logical) Whether to perform reliability test, default FALSE</p> <p>reliability_test.nfold (integer) Cross-validation folds for reliability test, default 10</p> <p>reliability_test.n (integer) Number of cells to use for reliability test, default 10</p> <p>cell_evaluation (logical) Whether to perform cell evaluation, default FALSE</p> <p>cell_evaluation.benchmark_data .RData Benchmark data for cell evaluation, default NULL</p> <p>cell_evaluation.FDR (numeric) FDR threshold for cell evaluation, default 0.05</p> <p>cell_evaluation.bootstrap_n (integer) Number of bootstrap samples for cell evaluation, default 10</p> <p>scPP ref_group (integer or character) Reference group or baseline for binary comparisons, e.g. "Normal" for Tumor/Normal studies and 0 for 0/1 case-control studies. default: 0</p>

Log2FC_cutoff (numeric) Minimum log2 fold-change for binary markers, default 0.585

estimate_cutoff (numeric) Effect size threshold for continuous traits, default 0.2

probs (numeric) Quantile cutoff for cell classification, default 0.2

scPAS assay (character) Assay to use from sc_data, default "RNA"

imputation (logical) Whether to perform imputation, default FALSE

nfeature (integer) Number of features to select, default 3000

alpha (numeric or NULL) Significance threshold, When NULL, alpha will keep increasing iteratively until the corresponding cells are screened out, default 0.01

independent (logical) The background distribution of risk scores is constructed independently of each cell. default: TRUE

network_class (character) Network class to use. default: 'SC', indicating gene-gene similarity networks derived from single-cell data. The other one is 'bulk'.

permutation_times (integer) Number of permutations, default 2000

FDR_threshold (numeric) FDR value threshold for identifying phenotype-associated cells default 0.05

scAB alpha (numeric) Coefficient of phenotype regularization ,default 0.005

alpha_2 (numeric) Coefficient of cell-cell similarity regularization, default 0.005

maxiter (integer) NMF optimization iterations, default 2000

tred (integer) Z-score threshold, default 2

DEGAS sc_data.pheno_colname (character) Phenotype column name in sc_data, default "NULL"

select_fraction (numeric) Fraction of cells to select for DEGAS, default 0.05

tmp_dir (character) Temporary directory for DEGAS, default "NULL"

env_params (list) Environment parameters for DEGAS, default "list()"

degas_params (list) DEGAS parameters, default "list()"

normality_test_method (character) Normality test method for DEGAS, default "jarque-bera"

LP_SGL resolution (numeric) Resolution parameter for Leiden clustering, default 0.6

alpha (numeric) Alpha parameter for SGL balancing L1 and L2 penalties, default 0.5

nfold (integer) Number of folds for cross-validation, default 5

dge_analysis (list) Differential expression analysis settings:

- run: (logical) Whether to run DEG analysis, default FALSE
- logFC_threshold: (numeric) Log fold change threshold, default 1
- pval_threshold: (numeric) P-value threshold, default 0.05

Value

A list containing:

scRNA_data A Seurat object with phenotype-associated cells labelled in `meta.data` column

Some screen_result Important information about the screened result related to the selected method

Data Matching Requirements

- `matched_bulk` column names and phenotype names/rownames must be identical
- Phenotype values must correspond to bulk samples (not directly to single cells)
- Mismatches will trigger an error before analysis begins, and there is a built-in pre-run check.

Method Compatibility

Method	Supported Phenotypes	Additional Parameters
Scissor	All three types	<code>alpha</code> , <code>cutoff</code> , <code>path2load_scissor_cache</code> , <code>path2save_scissor_inputs</code> , <code>reliability_test</code> , <code>reliability_threshold</code>
scPP	All three types	<code>ref_group</code> , <code>Log2FC_cutoff</code> , <code>estimate_cutoff</code> , <code>probs</code>
scPAS	All three types	<code>n_components</code> , <code>assay</code> , <code>imputation</code> , <code>nfeature</code> , <code>alpha</code> , <code>network_class</code> , <code>permutation_times</code> , <code>FDR</code> , <code>normality_test</code>
scAB	Binary/Survival	<code>alpha</code> , <code>alpha_2</code> , <code>maxiter</code> , <code>tred</code>
DEGAS	All three types	<code>sc_data</code> , <code>pheno_colname</code> , <code>select_fraction</code> , <code>tmp_dir</code> , <code>env_params</code> , <code>degas_params</code> , <code>normality_test</code>
LP_SGL	All three types	<code>resolution</code> , <code>alpha</code> , <code>nfold</code> , <code>dge_analysis</code>

See Also

Associated functions:

- [DoScissor](#)
- [DoscPP](#)
- [DoscPAS](#)
- [DoscAB](#)
- [DoDEGAS](#)
- [DoLP_SGL](#)

Description

Generates stacked bar plots showing the proportion of cells classified as Positive/Negative/Neutral by single-cell screening algorithms (Scissor, scPAS, scPP, or scAB) across different sample groups. Supports both single and multiple screen types visualization.

Usage

```
ScreenFractionPlot(
  screened_seurat,
  group_by = "Source", # x-axis label
  screen_type = NULL, # default: search all available screens
  show_null = FALSE,
  plot_color = NULL, # stack bar color of each status
  show_plot = FALSE,
  plot_title = "Screen Fraction",
  stack_width = 0.85,
  x_text_angle = 45,
  axis_linewidth = 0.8,
  legend_position = "right",
  x_lab = NULL,
  y_lab = "Fraction of Status",
  ncol = 2, # number of columns for facet wrap
  nrow = NULL, # number of rows for facet wrap
  verbose = SigBridgeRUtils::getFuncOption("verbose"),
  ... # ggplot2 theme arguments
)
```

Arguments

screened_seurat	A Seurat object containing screening results in metadata. Must contain columns corresponding to screen_type.
group_by	Metadata column name for grouping samples (default: "Source").
screen_type	Screening algorithm(s) used. Can be a single value or vector. Must match meta-data column(s) (case-sensitive, e.g., "scissor" for Scissor results).
show_null	Logical whether to show groups with zero cells (default: FALSE).
plot_color	Custom color palette (named vector format): - Required names: "Positive", "Negative", "Neutral" - Default: c("Neutral"="#CECECE", "Other"="#CECECE", "Positive"="#ff3333", "Negative"="#386c9b")
show_plot	Logical to immediately display plot (default: TRUE).
plot_title	Plot title (default: "Screen Fraction"). When multiple screen types, can be a vector of titles or single title (will append screen type).

<code>stack_width</code>	Bar width (default: 0.85).
<code>x_text_angle</code>	X-axis label angle (default: 45).
<code>axis_linewidth</code>	Axis line thickness (default: 0.8).
<code>legend_position</code>	Legend position (default: "right").
<code>x_lab</code>	X-axis label (default: NULL).
<code>y_lab</code>	Y-axis label (default: "Fraction of Status").
<code>ncol</code>	Number of columns for facet wrap when multiple screen types (default: 2).
<code>nrow</code>	Number of rows for facet wrap when multiple screen types (default: NULL).
<code>verbose</code>	Logical, whether to print a message
<code>...</code>	Other arguments passed to <code>ggplot2::theme()</code>

Value

A list containing:

- `stats`: A data frame (single screen) or list of data frames (multiple screens) with screening statistics including:
 - Grouping variable counts
 - Raw cell counts
 - Percentage fractions
- `plot`: A `ggplot2` object (single screen) or list of `ggplot2` objects (multiple screens)
- `combined_plot`: A combined plot using `patchwork` (only for multiple screens)

Visualization Details

- Bars are ordered by descending Positive fraction
- Y-axis shows percentage (0-100%)
- Zero-fraction groups are automatically hidden unless `show_null=TRUE`
- For multiple screen types, plots can be combined using `patchwork`

See Also

Other visualization_function: [ScreenUpset\(\)](#)

Examples

```
## Not run:
# Single screen type usage
res <- ScreenFractionPlot(
  screened_seurat = scissor_result,
  group_by = "PatientID",
  screen_type = "scissor"
)

# Multiple screen types at once
```



```

multi_res <- ScreenFractionPlot(
  screened_seurat = multi_screened_result,
  group_by = "TissueType",
  screen_type = c("scissor", "scPAS", "scPP", "scAB", "DEGAS"),
  ncol = 2
)

## End(Not run)

```

ScreenUpset	<i>ScreenUpset - Visualize cell type intersections from screened Seurat object</i>
-------------	--

Description

This function creates an UpSet plot to visualize intersections between different screening methods (e.g., scissor, scAB, scPAS, scPP) in a Seurat object metadata. It calculates all possible combinations of screening types and displays the number of cells positive for each combination.

Usage

```

ScreenUpset(
  screened_seurat,
  screen_type = NULL,
  show_plot = FALSE,
  n_intersections = 20,
  x_lab = "Screen Set Intersections",
  y_lab = "Number of Cells",
  title = "Cell Counts Across Screen Set Intersections",
  bar_color = "#4E79A7",
  combmatrix_point_color = "black",
  verbose = SigBridgeRUtils::getFuncOption("verbose"),
  ...
)

```

Arguments

screened_seurat	A Seurat object containing screening results in metadata. Must contain columns with screening types marked as "Positive".
screen_type	Character vector of screening types to analyze. Default: NULL, indicating that a self-search pattern will be used.
show_plot	Whether to show the upset plot. Default: FALSE
n_intersections	Number of intersections to display in the plot. Default: 20.
x_lab	Label for the x-axis. Default: "Screen Set Intersections".

<code>y_lab</code>	Label for the y-axis. Default: "Number of Cells".
<code>title</code>	Plot title. Default: "Cell Counts Across Screen Set Intersections".
<code>bar_color</code>	Color for the bars in the plot. Default: "#4E79A7".
<code>combmatrix_point_color</code>	Color for points in the combination matrix. Default: "black".
<code>verbose</code>	Logical, whether to print a message
<code>...</code>	Additional arguments passed to <code>ggplot2::theme()</code> for customizing the plot appearance.

Details

The function performs the following steps:

1. Validates input parameters and checks if specified screening types exist in metadata
2. Generates all possible combinations of screening types (from 1 to the total number of types)
3. Creates a logical matrix of positive cells for efficient computation
4. Calculates cell counts for each intersection using vectorized operations
5. Creates an UpSet plot visualization with customizable appearance

Value

A list containing two elements:

- `plot`: A ggplot object displaying the UpSet plot
- `stats`: A data frame with intersection statistics including:
 - `intersection`: Name of the intersection
 - `sets`: List of screening types in the intersection
 - `count`: Number of cells positive for all screening types in the intersection

See Also

Other visualization_function: [ScreenFractionPlot\(\)](#)

Examples

```
## Not run:
# Basic usage with default parameters
result <- ScreenUpset(screened_seurat = my_seurat_obj)

# Customize screening types and appearance
result <- ScreenUpset(
  screened_seurat = my_seurat_obj,
  screen_type = c("scissor", "scAB"),
  n_intersections = 15,
  title = "Custom Title",
  bar_color = "darkred"
)
```

```
## End(Not run)
```

setFuncOption

Configuration Functions for SigBridgeR Package

Description

These functions provide a centralized configuration system for the SigBridgeR package, allowing users to set and retrieve package-specific options with automatic naming conventions.

setFuncOption sets one or more configuration options for the SigBridgeR package. Options are automatically prefixed with "SigBridgeR." if not already present, ensuring proper namespace isolation.

getFuncOption retrieves configuration options for the SigBridgeR package. The function automatically handles the "SigBridgeR." prefix, allowing users to reference options with or without the explicit prefix.

checkFuncOption validates configuration options for the SigBridgeR package. This function ensures that all options meet type and value requirements before they are set in the global options.

Usage

```
setFuncOption(...)
```

Arguments

... Named arguments representing option-value pairs. Options can be provided with or without the "SigBridgeR." prefix. If the prefix is missing, it will be automatically added.

Details

This function performs the following validations:

verbose, parallel Must be single logical values (TRUE/FALSE)

parallel.type Must be a single character string

workers, timeout, seed Must be single integer values

The function is automatically called by setFuncOption to ensure all configuration options are valid before they are set.

Value

Invisibly returns NULL. The function is called for its side effects of setting package options.

The value of the specified option if it exists, otherwise the provided default value.

No return value. The function throws an error if the value doesn't meet the required specifications for the given option.

Available Options

- `verbose`: A logical value indicating whether to print verbose messages, defaults to `TRUE`
- `timeout`: An integer specifying the timeout in seconds for parallel processing, defaults to `'180L'`
- `seed`: An integer specifying the random seed for reproducible results, defaults to `'123L'`

Examples

```
# Set options with automatic prefixing
setFuncOption(verbose = TRUE)

# Retrieve options with automatic prefixing
getFuncOption("verbose")
```

SetupPyEnv

Create or Use Python Environment with Required Packages

Description

Sets up a Python environment with specified packages for DEGAS screening methods. This function can create new environments or reuse existing ones, supporting both Conda and venv environment types. It ensures all required dependencies are properly installed and verified.

Default method for unsupported environment types. Throws an informative error with supported environment types.

Usage

```
SetupPyEnv(env_type = c("conda", "venv"), ...)
```

Arguments

<code>env_type</code>	Character string specifying the type of Python environment to create or use. One of: <code>"conda"</code> , <code>"venv"</code> .
<code>...</code>	Additional parameters passed to specific environment methods.

Details

This function provides a comprehensive solution for Python environment management in R projects, particularly for machine learning workflows requiring TensorFlow. Key features include:

- **Environment Creation**: Automatically creates new environments or reuses existing ones with the same name
- **Package Management**: Installs specified Python packages with version pinning support
- **Verification**: Validates environment setup and package installations
- **Flexible Methods**: Supports different backend methods for environment creation (reticulate vs system calls)

The function uses S3 method dispatch to handle different environment types, allowing for extensible support of additional environment managers in the future.

Value

A data frame containing verification results for the environment setup, including installation status of all required packages. Invisibly returns the verification results.

See Also

[reticulate::conda_create()], [reticulate::virtualenv_create()] for underlying environment creation functions.

Examples

```
## Not run:
# Setup a Conda environment with default parameters
SetupPyEnv("conda")

# Setup a venv environment
SetupPyEnv("venv")

## End(Not run)
```

SymbolConvert	<i>Convert Ensembles Version IDs & TCGA Version IDs to Genes in Bulk Expression Data</i>
---------------	--

Description

Preprocess bulk expression data: convert Ensembles version IDs and TCGA version IDs to genes. NA values are replaced with unknown_k format (k stands for the position of the NA value in the row).

Usage

```
SymbolConvert(data, unknown_format = "unknown_{k}")
```

Arguments

- data bulk expression data (matrix or data.frame)
- unknown_format A glue pattern containing {k} for replace the NA value during conversion. k must be wrapped in curly braces, stands for the position of the NA value in the row. Default: "unknown_{k}".

Index

- * **DEGAS**
 - DoDEGAS, [9](#)
- * **LP_SGL**
 - DoLP_SGL, [12](#)
- * **PIPET**
 - DoPIPET, [14](#)
- * **duplicate aggregation**
 - aggregate-dups, [5](#)
- * **scAB**
 - DoscAB, [16](#)
- * **scPAS**
 - DoscPAS, [20](#)
- * **scPP**
 - DoscPP, [22](#)
- * **scissor**
 - DoScissor, [17](#)
- * **screen_method**
 - DoDEGAS, [9](#)
 - DoLP_SGL, [12](#)
 - DoPIPET, [14](#)
 - DoscAB, [16](#)
 - DoScissor, [17](#)
 - DoscPAS, [20](#)
 - DoscPP, [22](#)
- * **single_cell_preprocess**
 - FindRobustElbow, [24](#)
 - QCPatternDetect, [31](#)
- * **visualization_function**
 - ScreenFractionPlot, [39](#)
 - ScreenUpset, [41](#)
- AddMetaFeature, [3](#)
- AddMisc, [4](#)
- aggregate-dups, [5](#)
- AggregateDupCols, [5](#)
- AggregateDupCols (aggregate-dups), [5](#)
- AggregateDupRows, [5](#)
- AggregateDupRows (aggregate-dups), [5](#)
- AggregateDups, [5](#)
- AggregateDups (aggregate-dups), [5](#)
- BulkPreProcess, [6](#)
- cor, [9](#)
- cpm, [9](#)
- CreateSeuratObject, [34](#)
- DEGASEnvSet, [11](#)
- DEGASFindPy, [11](#)
- DEGASModelDetect, [11](#)
- DEGASParamSet, [11](#)
- DoDEGAS, [9](#), [13](#), [15](#), [17](#), [19](#), [21](#), [23](#), [38](#)
- DoLP_SGL, [11](#), [12](#), [15](#), [17](#), [19](#), [21](#), [23](#), [38](#)
- DoPIPET, [11](#), [13](#), [14](#), [17](#), [19](#), [21](#), [23](#)
- DoscAB, [11](#), [13](#), [15](#), [16](#), [19](#), [21](#), [23](#), [38](#)
- DoScissor, [11](#), [13](#), [15](#), [17](#), [17](#), [21](#), [23](#), [38](#)
- DoScissorCellEval, [19](#)
- DoScissorRelTest, [19](#)
- DoscPAS, [11](#), [13](#), [15](#), [17](#), [19](#), [20](#), [23](#), [38](#)
- DoscPP, [11](#), [13](#), [15](#), [17](#), [19](#), [21](#), [22](#), [38](#)
- FilterTumorCell, [24](#), [31](#)
- FindClusters, [34](#)
- FindNeighbors, [34](#)
- FindRobustElbow, [24](#), [31](#)
- FindVariableFeatures, [34](#)
- getFuncOption, [25](#)
- ListPyEnv, [26](#)
- LoadRefData, [28](#)
- MergeResult, [29](#)
- NormalizeData, [34](#)
- Pattern2Colname, [24](#), [31](#)
- PercentageFeatureSet, [31](#)
- prcomp, [9](#)
- ProcessSeuratObject, [24](#), [31](#)
- QCFilter, [30](#)

QCPatternDetect, [24](#), [30](#), [31](#)

RunPCA, [34](#)

RunTSNE, [34](#)

RunUMAP, [34](#)

ScaleData, [34](#)

SCPreProcess, [32](#)

Screen, [35](#)

ScreenFractionPlot, [39](#), [42](#)

ScreenUpset, [40](#), [41](#)

setFuncOption, [43](#)

SetupPyEnv, [44](#)

SymbolConvert, [9](#), [45](#)