# Package ‘scGate’

**Type** Package  
**Title** Marker-Based Cell Type Purification for Single-Cell Sequencing  
**Data**  
**Version** 1.4.1  
**Description**  
A common bioinformatics task in single-cell data analysis is to purify a cell type or cell population of interest from heterogeneous datasets. 'scGate' automatizes marker-based purification of specific cell populations, without requiring training data or reference gene expression profiles. Briefly, 'scGate' takes as input: i) a gene expression matrix stored in a 'Seurat' object and ii) a "gating model" (GM), consisting of a set of marker genes that define the cell population of interest. The GM can be as simple as a single marker gene, or a combination of positive and negative markers. More complex GMs can be constructed in a hierarchical fashion, akin to gating strategies employed in flow cytometry. 'scGate' evaluates the strength of signature marker expression in each cell using the rank-based method 'UCell', and then performs k-nearest neighbor (kNN) smoothing by calculating the mean 'UCell' score across neighboring cells. kNN-smoothing aims at compensating for the large degree of sparsity in scRNA-seq data. Finally, a universal threshold over kNN-smoothed signature scores is applied in binary decision trees generated from the user-provided gating model, to annotate cells as either “pure” or “impure”, with respect to the cell population of interest. See the related publication Andreatta et al. (2022) [doi:10.1093/bioinformatics/btac141].

**biocViews**
**Depends** R(>= 4.2.0)  
**Imports** Seurat(>= 4.0.0), UCell(>= 2.1.3), dplyr, stats, utils, methods, patchwork, ggridges, reshape2, ggplot2, BiocParallel  
**Suggests** ggparty, partykit, knitr, rmarkdown  
**VignetteBuilder** knitr  
**URL** [https://github.com/carmonalab/scGate](https://github.com/carmonalab/scGate)  
**BugReports** [https://github.com/carmonalab/scGate/issues](https://github.com/carmonalab/scGate/issues)  
**License** GPL-3  
**Encoding** UTF-8  
**LazyData** true
combine_scGate_multiclass

**Description**

If a single-cell dataset has precomputed results for multiple scGate models, combined them in multi-class annotation

**Usage**

```r
combine_scGate_multiclass(
  obj,
  prefix = "is.pure_",
  scGate_classes = NULL,
  min_cells = 20,
  multi.asNA = FALSE,
  out_column = "scGate_multi"
)
```
### gating_model

#### Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>obj</td>
<td>Seurat object with scGate results for multiple models stored as metadata</td>
</tr>
<tr>
<td>prefix</td>
<td>Prefix in metadata column names for scGate result models</td>
</tr>
<tr>
<td>scGate_classes</td>
<td>Vector of scGate model names. If NULL, use all columns that start with &quot;prefix&quot; above.</td>
</tr>
<tr>
<td>min_cells</td>
<td>Minimum number of cells for a cell label to be considered</td>
</tr>
<tr>
<td>multi.asNA</td>
<td>How to label cells that are &quot;Pure&quot; for multiple annotations: &quot;Multi&quot; (FALSE) or NA (TRUE)</td>
</tr>
<tr>
<td>out_column</td>
<td>The name of the metadata column where to store the multi-class cell labels</td>
</tr>
</tbody>
</table>

#### Value

A Seurat object with multi-class annotations based on the combination of multiple models. A new column (by default "scGate_multi") is added to the metadata of the Seurat object.

#### Examples

```r
# Define gating models
model.B <- gating_model(name = "Bcell", signature = c("MS4A1"))
model.T <- gating_model(name = "Tcell", signature = c("CD2","CD3D","CD3E"))

# Apply scGate with these models
query.seurat <- scGate(query.seurat, model=model.T, reduction="pca", output.col.name = "is.pure_Tcell")
query.seurat <- scGate(query.seurat, model=model.B, reduction="pca", output.col.name = "is.pure_Bcell")
query.seurat <- combine_scGate_multiclass(query.seurat, scGate_class=c("Tcell","Bcell"))
table(query.seurat$scGate_multi)
```

---

#### Description

Generate an scGate model from scratch or edit an existing one.

#### Usage

```r
gating_model(
  model = NULL,
  level = 1,
  name,
  signature,
  positive = TRUE,
)```
negative = FALSE,
remove = FALSE
)

Arguments

model  scGate model to be modified. When is NULL (default) a new model will be initialized.
level  integer. It refers to the hierarchical level of the model tree in which the signature will be added (level=1 by default)
name  Arbitrary signature name (i.e. Immune, Tcell, NK etc).
signature  character vector indicating gene symbols to be included in the signature (e.g. CD3D). If a minus sign is placed to the end of a gene name (e.g. "CD3D-"), this gene will be used as negative in UCell computing. See UCell documentation for details
positive  Logical indicating if the signature must be used as a positive signature in those model level. Default is TRUE.
negative  Same as ‘positive‘ but negated (negative=TRUE equals to positive=FALSE)
remove  Whether to remove the given signature from the model

Value

A scGate model that can be used by scGate to filter target cell types.

Examples

# create a simple gating model
my_model <- gating_model(level = 1, name = "immune", signature = c("PTPRC"))
my_model <- gating_model(model = my_model, level = 1, positive = FALSE,
     name = "Epithelial", signature = c("CDH1","FLT1") )
# Remove an existing signature
dropped_model <- gating_model(model = my_model, remove =TRUE, level = 1, name = "Epithelial")

genes.blacklist.default

Blocklist of genes for dimensionality reduction

Description

A list of signatures, for mouse and human. These include cell cycling, heat-shock genes, mitochondrial genes, and other genes classes, that may confound the identification of cell types. These are used internally by scGate and excluded from the calculation of dimensional reductions (PCA).

Format

A list of signatures
get_scGateDB

Load scGate model database

Description

Download, update or load local version of the scGate model database. These are stored in a GitHub repository, from where you can download specific versions of the database.

Usage

```r
get_scGateDB(
  destination = tempdir(),
  force_update = FALSE,
  version = "latest",
  branch = c("master", "dev"),
  verbose = FALSE,
  repo_url = "https://github.com/carmonalab/scGate_models"
)
```

Arguments

- `destination`: Destination path for storing the DB. The default is `tempdir()`; if you wish to edit locally the models and link them to the current project, set this parameter to a new directory name, e.g. `scGateDB`
- `force_update`: Whether to update an existing database.
- `version`: Specify the version of the scGate_models database (e.g. 'v0.1'). By default downloads the latest available version.
- `branch`: branch of the scGate model repository, either 'master' (default) or 'dev' for the latest models
- `verbose`: display progress messages
- `repo_url`: URL path to scGate model repository database

Details

Models for scGate are dataframes where each line is a signature for a given filtering level. A database of models can be downloaded using the function `get_scGateDB`. You may directly use the models from the database, or edit one of these models to generate your own custom gating model.

Value

A list of models, organized according to the folder structure of the database. See the examples below.

See Also

`scGate load_scGate_model`
Examples

```r
scGate.model.db <- get_scGateDB()
# To see a specific model, browse the list of models:
scGate.model.db$human$generic$Myeloid
# Apply scGate with this model
data(query.seurat)
query <- scGate(query.seurat, model=scGate.model.db$human$generic$Myeloid, reduction="pca")
```

---

### get_testing_data

**Download sample data**

**Description**

Helper function to obtain some sample data

**Usage**

```r
get_testing_data(version = "hsa.latest", destination = tempdir())
```

**Arguments**

- `version` : Which sample dataset
- `destination` : Save to this directory

**Value**

A list of datasets that can be used to test scGate

**Examples**

```r
testing.datasets <- get_testing_data(version = "hsa.latest")
```

---

### load_scGate_model

**Load a single scGate model**

**Description**

Load a custom scGate model into R. For the format of these models, have a look or edit one of the default models obtained with `get_scGateDB`

**Usage**

```r
load_scGate_model(model_file, master.table = "master_table.tsv")
```
performance.metrics

Arguments

model_file  scGate model file, in .tsv format.
master.table  File name of the master table (in repo_path folder) that contains cell type signatures.

Value

A scGate model in dataframe format, which can be given as input to the scGate function.

See Also

scGate get_scGateDB

Examples

dir <- tempdir() # this may also be set to your working directory
models <- get_scGateDB(destination=dir)
# Original or edited model
model.path <- paste0(dir,"/scGate_models-master/human/generic/Bcell_scGate_Model.tsv")
master.path <- paste0(dir,"/scGate_models-master/human/generic/master_table.tsv")
my.model <- load_scGate_model(model.path, master.path)
my.model

performance.metrics  Performance metrics

Description

Evaluate model performance for binary tasks

Usage

performance.metrics(actual, pred, return_contingency = FALSE)

Arguments

actual  Logical or numeric binary vector giving the actual cell labels.
pred  Logical or numeric binary vector giving the predicted cell labels.
return_contingency  Logical indicating if contingency table must be returned.

Value

Prediction performance metrics (Precision, Recall, MCC) between actual and predicted cell type labels.
Examples

```r
results <- performance.metrics(actual= sample(c(1,0),20,replace=TRUE),
    pred = sample(c(1,0),20,replace=TRUE,prob = c(0.65,0.35) )
)
```

plot_levels

Plot scGate filtering results by level

Description

Fast plotting of gating results over each model level.

Usage

```r
plot_levels(obj, pure.col = "green", impure.col = "gray")
```

Arguments

- `obj`: Gated Seurat object output of scGate filtering function
- `pure.col`: Color code for pure category
- `impure.col`: Color code for impure category

Value

UMAP plots with ‘Pure’/’Impure’ labels for each level of the scGate model

Examples

```r
scGate.model.db <- get_scGateDB()
model <- scGate.model.db$human$generic$Myeloid
# Apply scGate with this model
data(query.seurat)
query.seurat <- scGate(query.seurat, model=model,
     reduction="pca", save.levels=TRUE)
library(patchwork)
pll <- plot_levels(query.seurat)
wrap_plots(pll)
```
plot_tree

Plot model tree

Description

View scGate model as a decision tree (require ggparty package)

Usage

plot_tree(model, box.size = 8, edge.text.size = 4)

Arguments

- model: A scGate model to be visualized
- box.size: Box size
- edge.text.size: Edge text size

Value

A plot of the model as a decision tree. At each level, green boxes indicate the 'positive' (accepted) cell types, red boxed indicate the 'negative' cell types (filtered out). The final Pure population is the bottom right subset in the tree.

Examples

library(ggparty)
models <- get_scGateDB()
plot_tree(models$human$generic$Tcell)

plot_UCell_scores

Plot UCell scores by level

Description

Show distribution of UCell scores for each level of a given scGate model

Usage

plot_UCell_scores(
  obj,
  model,
  overlay = 5,
  pos.thr = 0.2,
  neg.thr = 0.2,
  ncol = NULL,
  combine = TRUE
)
Arguments

- obj: Gated Seurat object (output of scGate)
- model: scGate model used to identify a target population in obj
- overlay: Degree of overlay for ggridges
- pos.thr: Threshold for positive signatures used in scGate model (set to NULL to disable)
- neg.thr: Threshold for negative signatures used in scGate model (set to NULL to disable)
- ncol: Number of columns in output object (passed to wrap_plots)
- combine: Whether to combine plots into a single object, or to return a list of plots

Value

Returns a density plot of UCell scores for the signatures in the scGate model, for each level of the model.

Either a plot combined by patchwork (combine=T) or a list of plots (combine=F)

Examples

```r
scGate.model.db <- get_scGateDB()
model <- scGate.model.db$human$generic$Tcell
# Apply scGate with this model
data(query.seurat)
query.seurat <- scGate(query.seurat, model=model, 
  reduction="pca", save.levels=TRUE)
# View UCell score distribution
plot_UCell_scores(query.seurat, model)
```

query.seurat

Toy dataset to test the package

Description

A downsampled version (300 cells) of the single-cell dataset by Zilionis et al. (2019) <doi:10.1016/j.immuni.2019.03.009>, with precalculated PCA and UMAP reductions.

Format

A Seurat object
scGate

Filter single-cell data by cell type

Description

Apply scGate to filter specific cell types in a query dataset

Usage

scGate(
  data,
  model,
  pos.thr = 0.2,
  neg.thr = 0.2,
  assay = NULL,
  slot = "data",
  ncores = 1,
  seed = 123,
  keep.ranks = FALSE,
  reduction = c("calculate", "pca", "umap", "harmony", "Liors_elephant"),
  min.cells = 30,
  nfeatures = 2000,
  pca.dim = 30,
  param_decay = 0.25,
  maxRank = 1500,
  output.col.name = "is.pure",
  k.param = 30,
  genes.blacklist = "default",
  multi.asNA = FALSE,
  additional.signatures = NULL,
  save.levels = FALSE,
  verbose = FALSE
)

Arguments

data Seurat object containing a query data set - filtering will be applied to this object
model A single scGate model, or a list of scGate models. See Details for this format
pos.thr Minimum UCell score value for positive signatures
neg.thr Maximum UCell score value for negative signatures
assay Seurat assay to use
slot Data slot in Seurat object
ncores Number of processors for parallel processing
seed Integer seed for random number generator
keep.ranks Store UCell rankings in Seurat object. This will speed up calculations if the same object is applied again with new signatures.

reduction Dimensionality reduction to use for knn smoothing. By default, calculates a new reduction based on the given assay; otherwise you may specify a precalculated dimensionality reduction (e.g. in the case of an integrated dataset after batch-effect correction)

min.cells Minimum number of cells to cluster or define cell types

nfeatures Number of variable genes for dimensionality reduction

pca.dim Number of principal components for dimensionality reduction

param_decay Controls decrease in parameter complexity at each iteration, between 0 and 1. param_decay == 0 gives no decay, increasingly higher param_decay gives increasingly stronger decay

maxRank Maximum number of genes that UCell will rank per cell

output.col.name Column name with 'pure/impure' annotation

k.param Number of nearest neighbors for knn smoothing

genes.blacklist Genes blacklisted from variable features. The default loads the list of genes in scGate::genes.blacklist.default; you may deactivate blacklisting by setting genes.blacklist=NULL

multi.asNA How to label cells that are "Pure" for multiple annotations: "Multi" (FALSE) or NA (TRUE)

additional.signatures A list of additional signatures, not included in the model, to be evaluated (e.g. a cycling signature). The scores for this list of signatures will be returned but not used for filtering.

save.levels Whether to save in metadata the filtering output for each gating model level

verbose Verbose output

Details

Models for scGate are data frames where each line is a signature for a given filtering level. A database of models can be downloaded using the function get_scGateDB. You may directly use the models from the database, or edit one of these models to generate your own custom gating model.

Multiple models can also be evaluated at once, by running scGate with a list of models. Gating for each individual model is returned as metadata, with a consensus annotation stored in scGate_multi metadata field. This allows using scGate as a multi-class classifier, where only cells that are "Pure" for a single model are assigned a label, cells that are "Pure" for more than one gating model are labeled as "Multi", all others cells are annotated as NA.

Value

A new metadata column is.pure is added to the query Seurat object, indicating which cells passed the scGate filter. The active.ident is also set to this variable.
test_my_model

See Also

load_scGate_model get_scGateDB plot_tree

Examples

### Test using a small toy set

data(query.seurat)
my_scGate_model <- gating_model(name = "Bcell", signature = c("MS4A1"))
query.seurat <- scGate(query.seurat, model = my_scGate_model, reduction="pca")
table(query.seurat$is.pure)

### Test with larger datasets

library(Seurat)

seurat_object <- testing.datasets[["JerbyArnon"]]
models <- get_scGateDB()

seurat_object <- scGate(seurat_object, model=models$human$generic$PanBcell)

seurat_object_filtered <- subset(seurat_object, subset=is.pure="Pure")

### Run multiple models at once

models <- get_scGateDB()

model.list <- list("Bcell" = models$human$generic$Bcell,
                   "Tcell" = models$human$generic$Tcell)

seurat_object <- scGate(seurat_object, model=model.list)

DimPlot(seurat_object, group.by = "scGate_multi")

---

test_my_model  Test your model

Description

Wrapper for fast model testing on 3 sampled datasets

Usage

test_my_model(
  model,
  testing.version = "hsa.latest",
  custom.dataset = NULL,
  target = NULL,
  plot = TRUE
)
test_my_model

Arguments

- **model**
  - scGate model in data.frame format

- **testing.version**
  - Character indicating the version of testing datasets to be used. By default "hsa-latest" will be used. It will be ignored if a custom dataset is provided (in Seurat format).

- **custom.dataset**
  - Seurat object to be used as a testing dataset. For testing purposes, metadata seurat object must contain a column named 'cell_type' to be used as a gold standard. Also a set of positive targets must be provided in the target variable.

- **target**
  - Positive target cell types. If default testing version is used this variable must be a character indicating one of the available target models ('immune', 'Lymphoid', 'Myeloid', 'Tcell', 'Bcell', 'CD8T', 'CD4T', 'NK', 'MoMacDC', 'Plasma_cell', 'PanBcell'). If a custom dataset is provided in Seurat format, this variable must be a vector of positive cell types in your data. The last case also require that such labels were named as in your cell_type meta.data column.

- **plot**
  - Whether to return plots to device

Value

Returns performance metrics for the benchmarking datasets, and optionally plots of the predicted cell type labels in reduced dimensionality space.

Examples

```r
scGate.model.db <- get_scGateDB()
# Browse the list of models and select one:
model.panBcell <- scGate.model.db$human$generic$PanBcell
# Test the model with available testing datasets
panBcell.performance <- test_my_model(model.panBcell, target = "PanBcell")
model.Myeloid <- scGate.model.db$human$generic$Myeloid
myeloid.performance <- test_my_model(model.Myeloid, target = "Myeloid")
```
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