Package 'GrabSVG'

December 6, 2023

Title Granularity-Based Spatially Variable Genes Identifications
Version 0.0.2
Description Identifying spatially variable genes is critical in linking molecular cell functions with tissue phenotypes. This package implemented a granularity-based dimension-agnostic tool for the identification of spatially variable genes. The detailed description of this method is available at Wang, J. and Li, J. et al. 2023 (Wang, J. and Li, J. (2023), <doi:10.1038 s41467-023-43256-5="">).</doi:10.1038>
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2 GrabSVG

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A Granularity-Based Approach to identify Spatially Variable Genes

Description

This function is designed to identify spatially variable genes through a granularity-based approach.

Usage

```
GrabSVG(Coords, ExpMat_Sp, D_1 = 1.0, D_2 = 3.0,
Exp_Norm = TRUE, Coords_Norm_Method = c("Sliced", "Overall", "None"))
```

Arguments

Coords A M x D matrix representing D-dimensional coordinates for M spots

ExpMat_Sp A sparse, N x M expression matrix in dgCMatrix class with N genes and M spots

D_1 Size of the small patch

D_2 Size of the big patch

Exp_Norm A Boolean value indicating whether the expression matrix should be normalized Coords_Norm_Method

Normalization method for the coordinates matrix, which can be "None" "Sliced"

Normalization method for the coordinates matrix, which can be "None", "Sliced", or "Overall".

Details

This function utilizes a MxD matrix (Coords) representing D-dimensional coordinates with M spots and a sparse, NxM expression matrix (ExpMat_Sp) with N genes and M spots.

Value

A data frame with the name of genes and corresponding p-values.

Examples

```
Coords <- expand.grid(1:100,1:100, 1:3)

RandFunc <- function(n) floor(10 * stats::rbeta(n, 1, 5))

Raw_Exp <- Matrix::rsparsematrix(nrow = 10^4, ncol = 3*10^4, density = 0.0001, rand.x = RandFunc)

Filtered_ExpMat <- SpFilter(Raw_Exp)

rownames(Filtered_ExpMat) <- paste0("Gene_", 1:nrow(Filtered_ExpMat))

P_values <- GrabSVG(Coords, Filtered_ExpMat)
```

LoadSpatial 3

LoadSpatial Loading data from a Seurat object or a data frame.
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Description

A function to load and filter data from a Seurat object or a data frame.

Usage

```
LoadSpatial(InputData, Dimension = 2)
```

Arguments

InputData A Seurat spatial object or a $M \times (D + N)$ data matrix representing the D-dimensional

coordinates and expressions of N genes on M spots. The coordinates should be

placed at the first D columns

Dimension The dimension of coordinates

Value

A list of two data frame:

Coords A M x D matrix representing D-dimensional coordinates for M spots

ExpMatrix A sparse, N x M expression matrix in dgCMatrix class with N genes and M

spots

SpFilter	A function for filtering low expressed genes	

Description

A function for filtering low expressed genes

Usage

```
SpFilter(ExpMat_Sp, Threshold = 5)
```

Arguments

ExpMat_Sp A sparse, N x M expression matrix in dgCMatrix class with N genes and M

spots

Threshold A threshold set to filter out genes with a total read count below this specified

value

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Value

A sparse expression matrix in dgCMatrix class

Examples

```
# create a sparse expression matrix
Raw_ExpMat <- Matrix::rsparsematrix(nrow = 10000, ncol = 2000,
density = 0.01, rand.x = function(n) rpois(n, 15))
Filtered_ExpMat <- SpFilter(Raw_ExpMat)</pre>
```

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