

Package: GrabSVG (via r-universe)

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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](https://doi.org/10.1038/s41467-023-43256-5)>).

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Encoding UTF-8

Imports Matrix, sparseMatrixStats, fitdistrplus, RANN, spam

Suggests knitr, rmarkdown

RoxygenNote 7.2.1

NeedsCompilation no

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Repository <https://castleli.r-universe.dev>

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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", 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	<i>Loading data from a Seurat object or a data frame.</i>
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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 $M \times D$ matrix representing D -dimensional coordinates for M spots
ExpMatrix	A sparse, $N \times M$ expression matrix in dgCMatrix class with N genes and M spots

SpFilter	<i>A function for filtering low expressed genes</i>
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Description

A function for filtering low expressed genes

Usage

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

Arguments

ExpMat_Sp	A sparse, $N \times M$ 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

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)
```

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