# Package 'Giotto'

December 13, 2019

```
Title Spatial single-cell transcriptomics pipeline.
Version 0.1.4
Description Pipeline to process, analyze and visualize (spatial) single-cell expression data.
License MIT + file LICENSE
Encoding UTF-8
LazyData true
RoxygenNote 7.0.1
Depends data.table (>= 1.12.2),
      ggplot2 (>= 3.1.1),
      base (>= 3.5.1),
      utils (>= 3.5.1),
      R (>= 3.5.1)
Imports Rtsne (>= 0.15),
      uwot (>= 0.0.0.9010),
      multinet (>= 3.0.2),
      FactoMineR (>= 1.34),
      factoextra (>= 1.0.5),
      cowplot (>= 0.9.4),
      grDevices,
      RColorBrewer (>= 1.1-2),
      jackstraw (>= 1.3),
      dbscan (>= 1.1-3),
      ggalluvial (>= 0.9.1),
      scales (>= 1.0.0),
      ComplexHeatmap (>= 1.20.0),
      qvalue (>= 2.14.1),
      lfa (>= 1.12.0),
      igraph (>= 1.2.4.1),
      plotly,
      reticulate,
      magrittr,
      limma,
      ggdendro,
      smfishHmrf,
      matrixStats (>= 0.55.0),
      IRanges,
      devtools,
```

reshape2

2 R topics documented:

```
Suggests knitr,
rmarkdown,
MAST,
scran (>= 1.10.1),
png,
tiff,
biomaRt
```

# biocViews

VignetteBuilder knitr

Remotes lambdamoses/smfishhmrf-r

# ${\sf R}$ topics documented:

addCellMetadata
addCellStatistics
addGeneMetadata
addGeneStatistics
addHMRF
addNetworkLayout
addStatistics
adjustGiottoMatrix
aes_string2
allCellCellcommunicationsScores
all_plots_save_function
annotateGiotto
annotateSpatialNetwork
annotate_spatlocs_with_spatgrid_2D
annotate_spatlocs_with_spatgrid_3D
average_gene_gene_expression_in_groups
binGetSpatialGenes
calculateHVG
calculateMetaTable
calculateMetaTableCells
calculate_spatial_genes_python
cellProximityBarplot
cellProximityEnrichment
cellProximityHeatmap
cellProximityNetwork
cellProximitySpatPlot
cellProximitySpatPlot2D
cellProximitySpatPlot3D
cellProximityVisPlot
cellProximityVisPlot_2D_ggplot
cellProximityVisPlot_2D_plotly
cellProximityVisPlot_3D_plotly
changeGiottoInstructions
clusterCells
clusterSpatialCorGenes
combineMetadata
convertEnsemblToGeneSymbol 46

createGiottoInstructions
createGiottoObject
createHeatmap_DT
createMetagenes
$create Nearest Network \dots \dots$
createSpatialEnrich
createSpatialGrid
createSpatialGrid_2D
createSpatialGrid_3D
createSpatialNetwork
create_average_detection_DT
create_average_DT
create_cell_type_random_cell_IDs
create_cluster_matrix
create_dimObject
decide_cluster_order
detectSpatialCorGenes
detectSpatialPatterns
dimCellPlot
dimCellPlot2D
dimGenePlot
dimGenePlot2D
dimGenePlot3D
dimPlot
dimPlot2D
dimPlot2D_single
dimPlot3D
direction_test_CPG
doHclust
doHMRF
doKmeans
dokineans
doLeidenSubCluster
doLouvainCluster
doLouvainCluster_community
doLouvainCluster_multinet
doLouvainSubCluster
doLouvainSubCluster_community
doLouvainSubCluster_multinet
doRandomWalkCluster
doSNNCluster
do_spatial_grid_averaging
do_spatial_knn_smoothing
dt_to_matrix
enrichSpatialCorGroups
exportGiottoViewer
$exprOnly Cell Cell communication Scores \\  \   \dots \\  \  \   \dots \\  \   \dots \\ \  \   \dots \\  \   \dots \\  \   \dots \\  \   \dots \\  \   \dots \\ \  \   \dots \\ \$
extended_gini_fun
extractNearestNetwork
fDataDT
filterCombinations
filterCPGscores

filterDistributions	113
filterGiotto	
findGiniMarkers	115
findGiniMarkers_one_vs_all	116
findMarkers	117
findMarkers_one_vs_all	119
findMastMarkers	120
findMastMarkers_one_vs_all	121
findScranMarkers	
findScranMarkers_one_vs_all	
find_grid_2D	
find_grid_3D	
find_grid_x	
find_grid_y	
find_grid_z	
fish_function	
fish_function2	
FSV_show	
GenePattern show	
general_save_function	
e = =	
get10Xmatrix	
getCellProximityGeneScores	
getClusterSimilarity	
getDendrogramSplits	
getDistinctColors	
getGeneToGeneScores	
get_cell_to_cell_sorted_name_conversion	
get_interaction_gene_enrichment	
get_specific_interaction_gene_enrichment	
ggplot_save_function	
giotto-class	
heatmSpatialCorGenes	
hyperGeometricEnrich	
kmeans_binarize	
loadHMRF	140
makeSignMatrixPAGE	141
makeSignMatrixRank	141
make simulated network	142
mergeClusters	142
mygini_fun	
nnDT_to_kNN	
node clusters	
normalizeGiotto	
OR_function2	
PAGEEnrich	
pDataDT	
plotCPGscores	
plotGTGscores	
plotHeatmap	
plotly_axis_scale_2D	
plotly_axis_scale_3D	
plotly_grid	154

plotly_network
plotMetaDataCellsHeatmap
plotMetaDataHeatmap
plotPCA
plotPCA_2D
plotPCA_3D
plotTSNE
plotTSNE_2D
plotTSNE_3D
plotUMAP
plotUMAP_2D
plotUMAP_3D
plot_network_layer_ggplot
plot_point_layer_ggplot
plot_spat_point_layer_ggplot
print.giotto
rankEnrich
rankSpatialCorGroups
rank_binarize
readGiottoInstructions
removeCellAnnotation
removeGeneAnnotation
replaceGiottoInstructions
runPCA
runtSNE
runUMAP
selectPatternGenes
select_expression_values
show,giotto-method
showClusterDendrogram
showClusterHeatmap
showCPGscores
showGeneExpressionProximityScore
showGiottoInstructions
showGTGscores
showIntExpressionProximityScore
showPattern
showPattern2D
showPattern3D
showPatternGenes
showProcessingSteps
showSpatialCorGenes
showTopGeneToGene
signPCA
spatCellPlot
spatCellPlot2D
spatDimCellPlot
spatDimCellPlot2D
spatDimGenePlot
spatDimGenePlot2D
spatDimGenePlot3D
spatDimPlot
- эринэнн ю

308

Index

spatDimPlot2D	28
spatDimPlot3D	2
spatGenePlot	
spatGenePlot2D	8
spatGenePlot3D	0
Spatial_AEH	1
Spatial_DE	2
spatPlot	3
spatPlot2D	6
spatPlot2D_single	.9
spatPlot3D	2
specificCellCellcommunicationScores	4
split_dendrogram_in_two	5
stitchFieldCoordinates	6
subClusterCells	7
subsetGiotto	9
subsetGiottoLocs	9
viewHMRFresults	0
viewHMRFresults2D	1
viewHMRFresults3D	2
violinPlot	3
visDimGenePlot	4
visDimGenePlot_2D_ggplot	6
visDimGenePlot_3D_plotly	7
visDimPlot	9
visDimPlot_2D_ggplot	1
visDimPlot_2D_plotly	'3
visDimPlot_3D_plotly	
visForceLayoutPlot	
visGenePlot	8
visGenePlot_2D_ggplot	0
visGenePlot_3D_plotly	
visPlot	
visPlot_2D_ggplot	
-	
visPlot_3D_plotly	
visSpatDimGenePlot	
visSpatDimGenePlot_2D	
visSpatDimGenePlot_3D	
visSpatDimPlot	
visSpatDimPlot_2D	
visSpatDimPlot_3D	
writeHMRFresults	
write_giotto_viewer_annotation	
write_giotto_viewer_dim_reduction	
write_giotto_viewer_numeric_annotation	
	1

addCellMetadata 7

addCellMetadata addCellMetadata

## **Description**

adds cell metadata to the giotto object

#### Usage

```
addCellMetadata(
  gobject,
  new_metadata,
  by_column = FALSE,
  column_cell_ID = NULL
)
```

# Arguments

gobject giotto object

new\_metadata new cell metadata to use (data.table, data.frame, ...)

by\_column merge metadata based on cell\_ID column in pDataDT (default = FALSE)

column\_cell\_ID column name of new metadata to use if by\_column = TRUE

# **Details**

You can add additional cell metadata in two manners: 1. Provide a data.table or data.frame with cell annotations in the same order as the cell\_ID column in pDataDT(gobject) 2. Provide a data.table or data.frame with cell annotations and specificy which column contains the cell IDs, these cell IDs need to match with the cell\_ID column in pDataDT(gobject)

## Value

giotto object

# **Examples**

addCellMetadata(gobject)

addCellStatistics addCellStatistics

# Description

adds cells statistics to the giotto object

8 addGeneMetadata

#### Usage

```
addCellStatistics(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  detection_threshold = 0,
  return_gobject = TRUE
)
```

# Arguments

## **Details**

This function will add the following statistics to cell metadata:

- nr\_genes: Denotes in how many genes are detected per cell
- perc\_genes: Denotes what percentage of genes is detected per cell
- total\_expr: Shows the total sum of gene expression per cell

## Value

```
giotto object if return_gobject = TRUE
```

## **Examples**

```
addCellStatistics(gobject)
```

addGeneMetadata

addGeneMetadata

#### **Description**

adds gene metadata to the giotto object

## Usage

```
addGeneMetadata(gobject, new_metadata, by_column = F, column_gene_ID = NULL)
```

```
gobject giotto object

new_metadata new metadata to use

by_column merge metadata based on gene_ID column in fDataDT

column_cell_ID column name of new metadata to use if by_column = TRUE
```

addGeneStatistics 9

#### **Details**

You can add additional gene metadata in two manners: 1. Provide a data.table or data.frame with gene annotations in the same order as the gene\_ID column in fDataDT(gobject) 2. Provide a data.table or data.frame with gene annotations and specificy which column contains the gene IDs, these gene IDs need to match with the gene\_ID column in fDataDT(gobject)

#### Value

giotto object

# **Examples**

addGeneMetadata(gobject)

addGeneStatistics

addGeneStatistics

#### **Description**

adds gene statistics to the giotto object

#### Usage

```
addGeneStatistics(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  detection_threshold = 0,
  return_gobject = TRUE
)
```

# **Arguments**

#### **Details**

This function will add the following statistics to gene metadata:

- nr\_cells: Denotes in how many cells the gene is detected
- per\_cells: Denotes in what percentage of cells the gene is detected
- total\_expr: Shows the total sum of gene expression in all cells
- mean\_expr: Average gene expression in all cells
- mean\_expr\_det: Average gene expression in cells with detectable levels of the gene

10 addHMRF

#### Value

```
giotto object if return_gobject = TRUE
```

# **Examples**

addGeneStatistics(gobject)

addHMRF

addHMRF

# Description

Add selected results from doHMRF to the giotto object

# Usage

```
addHMRF(gobject, HMRFoutput, k = NULL, betas_to_add = NULL, hmrf_name = NULL)
```

# Arguments

gobject giotto object

 $\label{eq:hmrf} HMRF \ output \ from \ do HMRF()$ 

k number of domains

name specify a custom name

# **Details**

Description ...

#### Value

giotto object

# **Examples**

addHMRF(gobject)

addNetworkLayout 11

addNetworkLayout

addNetworkLayout

# Description

Add a network layout for a selected nearest neighbor network

## Usage

```
addNetworkLayout(
  gobject,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  layout_type = c("drl"),
  options_list = NULL,
  layout_name = "layout",
  return_gobject = TRUE
)
```

# **Arguments**

# **Details**

This function creates layout coordinates based on the provided kNN or sNN. Currently only the force-directed graph layout "drl", see layout\_with\_drl, is implemented. This provides an alternative to tSNE or UMAP based visualizations.

# Value

giotto object with updated layout for selected NN network

# Examples

```
addNetworkLayout(gobject)
```

12 adjustGiottoMatrix

addStatistics

addStatistics

# Description

adds genes and cells statistics to the giotto object

## Usage

```
addStatistics(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  detection_threshold = 0,
  return_gobject = TRUE
)
```

# **Arguments**

## **Details**

See addGeneStatistics and addCellStatistics

# Value

```
giotto object if return_gobject = TRUE, else a list with results
```

# **Examples**

```
addStatistics(gobject)
```

adjustGiottoMatrix adjustGiottoMatrix

# Description

normalize and/or scale expresion values of Giotto object

aes\_string2

### Usage

```
adjustGiottoMatrix(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  batch_columns = NULL,
  covariate_columns = NULL,
  return_gobject = TRUE,
  update_slot = c("custom")
)
```

## **Arguments**

```
gobject giotto object

expression_values

expression values to use

batch_columns metadata columns that represent different batch (max = 2)

covariate_columns

metadata columns that represent covariates to regress out

return_gobject boolean: return giotto object (default = TRUE)

update_slot expression slot that will be updated (default = custom)
```

#### **Details**

This function implements the limma::removeBatchEffect function to remove known batch effects and to adjust expression values according to provided covariates.

#### Value

giotto object

# **Examples**

```
adjustGiottoMatrix(gobject)
```

```
aes_string2 aes_string2
```

# Description

makes sure aes\_string can also be used with names that start with numeric values

```
aes_string2(...)
```

```
all {\tt Cell Cell communications Scores} \\ all {\tt Cell Cell communications Scores}
```

## **Description**

All Cell-Cell communication scores based on spatial expression of interacting cells

# Usage

```
allCellCellcommunicationsScores(
  gobject,
  spatial_network_name = "spatial_network",
  cluster_column = "cell_types",
  random_iter = 100,
  gene_set_1,
  gene_set_2,
  log2FC_addendum = 0.1,
  min_observations = 2,
  verbose = c("a little", "a lot", "none")
)
```

# **Arguments**

```
gobject
                  giotto object to use
spatial_network_name
                  spatial network to use for identifying interacting cells
cluster_column cluster column with cell type information
random\_iter
                  number of iterations
                  first specific gene set from gene pairs
gene_set_1
                  second specific gene set from gene pairs
gene_set_2
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
                  verbose
verbose
```

## **Details**

Statistical framework to identify if pairs of genes (such as ligand-receptor combinations) are expressed at higher levels than expected based on a reshuffled null distribution of gene expression values in cells that are spatially in proximity to eachother. More details will follow soon.

## Value

Cell-Cell communication scores for gene pairs based on spatial interaction

# **Examples**

```
allCellCellcommunicationsScores(gobject)
```

```
all_plots_save_function 
 all_plots_save_function
```

# Description

Function to automatically save plots to directory of interest

# Usage

```
all_plots_save_function(
  gobject,
  plot_object,
  save_dir = NULL,
  save_folder = NULL,
  save_name = NULL,
  default_save_name = "giotto_plot",
  save_format = NULL,
  show_saved_plot = F,
  ncol = 1,
  nrow = 1,
  scale = 1,
  base_width = NULL,
  base_height = NULL,
  base_aspect_ratio = NULL,
  units = NULL,
  dpi = NULL,
  limitsize = TRUE,
)
```

```
gobject
                  giotto object
                  object to plot
plot_object
save_dir
                  directory to save to
save_folder
                  folder in save_dir to save to
save_name
                  name of plot
save_format
                  format (e.g. png, tiff, pdf, ...)
show_saved_plot
                  load & display the saved plot
ncol
                  number of columns
                  number of rows
nrow
scale
                  scale
base_width
                  width
base_height
                  height
{\tt base\_aspect\_ratio}
                  aspect ratio
```

16 annotateGiotto

```
units units

dpi Plot resolution

limitsize When TRUE (the default), ggsave will not save images larger than 50x50 inches, to prevent the common error of specifying dimensions in pixels.
```

additional parameters to ggplot\_save\_function or general\_save\_function

#### See Also

```
general_save_function
```

# **Examples**

```
all_plots_save_function(gobject)
```

annotateGiotto

annotateGiotto

# **Description**

Converts cluster results into provided annotation.

# Usage

```
annotateGiotto(
  gobject,
  annotation_vector = NULL,
  cluster_column = NULL,
  name = "cell_types"
)
```

#### **Arguments**

## Details

You need to specifify which (cluster) column you want to annotate and you need to provide an annotation vector like this:

- 1. identify the cell type of each cluster
- 2. create a vector of these cell types, e.g. cell\_types = c('T-cell', 'B-cell', 'Stromal')
- 3. provide original cluster names to previous vector, e.g. names(cell\_types) = c(2, 1, 3)

# Value

giotto object

## **Examples**

```
annotateGiotto(gobject)
```

 $annotate {\tt Spatial Network}$ 

annotateSpatialNetwork

# Description

Annotate spatial network with cell metadata information.

# Usage

```
annotateSpatialNetwork(
  gobject,
  spatial_network_name = "spatial_network",
  cluster_column
)
```

# **Arguments**

# Value

annotated network in data.table format

# **Examples**

```
annotateSpatialNetwork(gobject)
```

```
annotate\_spatlocs\_with\_spatgrid\_2D \\ annotate\_spatlocs\_with\_spatgrid\_2D
```

# Description

annotate spatial locations with 2D spatial grid information

# Usage

```
annotate_spatlocs_with_spatgrid_2D(spatloc, spatgrid)
```

```
spatloc spatial_locs slot from giotto object spatgrid selected spatial_grid slot from giotto object
```

#### Value

annotated spatial location data.table

# **Examples**

```
annotate_spatlocs_with_spatgrid_2D()
```

```
annotate\_spatlocs\_with\_spatgrid\_3D \\ annotate\_spatlocs\_with\_spatgrid\_3D
```

# Description

annotate spatial locations with 3D spatial grid information

# Usage

```
annotate_spatlocs_with_spatgrid_3D(spatloc, spatgrid)
```

# **Arguments**

```
spatloc spatial_locs slot from giotto object spatgrid selected spatial_grid slot from giotto object
```

## Value

annotated spatial location data.table

## **Examples**

```
annotate_spatlocs_with_spatgrid_3D()
```

```
average_gene_gene_expression_in_groups

average_gene_gene_expression_in_groups
```

# **Description**

calculate average expression per cluster

```
average_gene_gene_expression_in_groups(
  gobject,
  cluster_column = "cell_types",
  gene_set_1,
  gene_set_2
)
```

binGetSpatialGenes 19

# **Arguments**

```
gobject giotto object to use
cluster_column cluster column with cell type information
gene_set_1 first specific gene set from gene pairs
gene_set_2 second specific gene set from gene pairs
```

#### **Details**

Details will follow soon.

#### Value

data.table with average expression scores for each cluster

#### **Examples**

```
average_gene_gene_expression_in_groups(gobject)
```

binGetSpatialGenes

binGetSpatialGenes

#### **Description**

Rapid computation of genes that are spatially clustered

#### Usage

```
binGetSpatialGenes(
  gobject,
  bin_method = c("kmeans", "rank"),
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_network_name = "spatial_network",
  nstart = 3,
  iter_max = 10,
  percentage_rank = 10,
  do_fisher_test = F,
  community_expectation = 5,
  verbose = F
)
```

20 calculateHVG

nstart kmeans: nstart parameter iter\_max kmeans: iter.max parameter

do\_fisher\_test perform fisher test

community\_expectation

cell degree expectation in spatial communities

verbose be verbose

rank\_percentage

percentage of top cells for binarization

## **Details**

We provide two ways to identify spatial genes based on gene expression binarization. Both methods are identicial except for how binarization is performed.

- 1. binarize: Each gene is binarized (0 or 1) in each cell with **kmeans** (k = 2) or based on **rank** percentile
- 2. network: Alll cells are connected through a k-nearest neighbor network
- 3. contingency table: A contingency table is calculated based on all pairwise cell-cell interactions (0-0, 0-1, 1-0 or 1-1)
- 4. For each gene an odds-ratio (OR) and fisher.test (optional) is calculated

Additionally 2 other statistics are provided:

- Number of cells with high expression (binary = 1)
- total and ratio of highly connected cells: Cells with a connectivity higher than community\_expectation

By selecting a subset of likely spatial genes (e.g. highly variable genes) the function will be much faster.

## Value

data.table with results (see details)

# **Examples**

binGetSpatialGenes(gobject)

calculateHVG calculateHVG

# **Description**

compute highly variable genes

calculateHVG 21

#### Usage

```
calculateHVG(
      gobject,
      expression_values = c("normalized", "scaled", "custom"),
      method = c("cov_groups", "cov_loess"),
      reverse_log_scale = FALSE,
      logbase = 2,
      expression_threshold = 0,
      nr_expression_groups = 20,
      zscore_threshold = 1.5,
      HVGname = "hvg",
      difference_in_cov = 0.1,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "HVGplot",
      return_gobject = TRUE
    )
Arguments
    gobject
                     giotto object
    expression_values
                     expression values to use
                     method to calculate highly variable genes
    method
    reverse_log_scale
                     reverse log-scale of expression values (default = FALSE)
    logbase
                     if reverse_log_scale is TRUE, which log base was used?
    expression_threshold
                     expression threshold to consider a gene detected
    nr_expression_groups
                     number of expression groups for cov_groups
    zscore_threshold
                     zscore to select hvg for cov_groups
    HVGname
                     name for highly variable genes in cell metadata
    difference_in_cov
                     minimum difference in coefficient of variance required
    show_plot
                     show plot
                     return ggplot object
    return_plot
                     directly save the plot [boolean]
    save_plot
                     list of saving parameters from all_plots_save_function
    save_param
    default_save_name
                     default save name for saving, don't change, change save_name in save_param
```

return\_gobject boolean: return giotto object (default = TRUE)

22 calculateMetaTable

#### **Details**

Currently we provide 2 ways to calculate highly variable genes: 1. high coeff of variance (COV) within groups:

First genes are binned (*nr\_expression\_groups*) into average expression groups and the COV for each gene is converted into a z-score within each bin. Genes with a z-score higher than the threshold (*zscore\_threshold*) are considered highly variable.

# 2. high COV based on loess regression prediction:

A predicted COV is calculated for each gene using loess regression (COV~log(mean expression)) Genes that show a higher than predicted COV (*difference\_in\_cov*) are considered highly variable.

#### Value

giotto object highly variable genes appended to gene metadata (fDataDT)

#### **Examples**

```
calculateHVG(gobject)
```

calculateMetaTable

calculateMetaTable

#### **Description**

calculates the average gene expression for one or more (combined) annotation columns.

### Usage

```
calculateMetaTable(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  metadata_cols = NULL,
  selected_genes = NULL
)
```

#### **Arguments**

```
gobject giotto object
expression_values
expression values to use
metadata_cols annotation columns found in pDataDT(gobject)
selected_genes subset of genes to use
```

# Value

data.table with average expression values for each gene per (combined) annotation

## **Examples**

```
calculateMetaTable(gobject)
```

calculateMetaTableCells 23

```
calculateMetaTableCells
```

calculateMetaTableCells

## **Description**

calculates the average metadata values for one or more (combined) annotation columns.

#### Usage

```
calculateMetaTableCells(
  gobject,
  value_cols = NULL,
  metadata_cols = NULL,
  spat_enr_names = NULL
)
```

## **Arguments**

```
gobject giotto object
value_cols metadata or enrichment value columns to use
metadata_cols annotation columns found in pDataDT(gobject)
spat_enr_names which spatial enrichment results to include
```

#### Value

data.table with average metadata values per (combined) annotation

#### **Examples**

```
calculateMetaTableCells(gobject)
```

```
calculate\_spatial\_genes\_python \\ calculate\_spatial\_genes\_python
```

## **Description**

Calculate spatial genes using distance matrix.

```
calculate_spatial_genes_python(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  metric = "euclidean",
  subset_genes = NULL,
  rbp_p = 0.95,
  examine_top = 0.3,
  python_path = NULL
)
```

## **Arguments**

```
gobject giotto object
expression_values
expression values to use

metric distance metric to use
subset_genes only run on this subset of genes
rbp_p fractional binarization threshold
examine_top top fraction to evaluate with silhouette
python_path specify specific path to python if required
```

#### **Details**

Description of how we compute spatial pattern genes.

#### Value

data.table with spatial scores

## **Examples**

```
calculate_spatial_genes_python(gobject)
```

```
cellProximityBarplot cellProximityBarplot
```

# Description

Create barplot from cell-cell proximity scores

```
cellProximityBarplot(
  gobject,
  CPscore,
  min_orig_ints = 5,
  min_sim_ints = 5,
  p_val = 0.05,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "cellProximityBarplot"
)
```

#### **Arguments**

```
gobject
                  giotto object
                  CPscore, output from cellProximityEnrichment()
CPscore
                  filter on minimum original cell-cell interactions
min_orig_ints
min_sim_ints
                  filter on minimum simulated cell-cell interactions
p_val
                  p-value
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
```

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

## **Details**

This function creates a barplot that shows the spatial proximity enrichment or depletion of cell type pairs.

#### Value

ggplot barplot

# **Examples**

```
cellProximityBarplot(CPscore)
```

```
{\tt cellProximityEnrichment}
```

cellProximityEnrichment

# **Description**

Compute cell-cell interaction enrichment (observed vs expected)

## Usage

```
cellProximityEnrichment(
 gobject,
  spatial_network_name = "spatial_network",
 cluster_column,
  number_of_simulations = 100
)
```

```
gobject
                 giotto object
spatial_network_name
                 name of spatial network to use
cluster_column name of column to use for clusters
number_of_simulations
                 number of simulations to create expected observations
```

#### **Details**

Spatial proximity enrichment or depletion between pairs of cell types is calculated by calculating the observed over the expected frequency of cell-cell proximity interactions. The expected frequency is the average frequency calculated from a number of spatial network simulations. Each individual simulation is obtained by reshuffling the cell type labels of each node (cell) in the spatial network.

#### Value

List of cell Proximity scores (CPscores) in data.table format. The first data.table (raw\_sim\_table) shows the raw observations of both the original and simulated networks. The second data.table (enrichm\_res) shows the enrichment results.

# **Examples**

```
cellProximityEnrichment(gobject)
```

```
cellProximityHeatmap cellProximityHeatmap
```

#### **Description**

Create heatmap from cell-cell proximity scores

# Usage

```
cellProximityHeatmap(
  gobject,
  CPscore,
  scale = T,
  order_cell_types = T,
  color_breaks = NULL,
  color_names = NULL,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "cellProximityHeatmap")
```

```
gobject giotto object

CPscore CPscore, output from cellProximityEnrichment()

scale scale cell-cell proximity interaction scores

order_cell_types

order cell types based on enrichment correlation

color_breaks numerical vector of length 3 to represent min, mean and maximum

color_names character color vector of length 3

show_plot show plot
```

cellProximityNetwork 27

#### **Details**

This function creates a heatmap that shows the spatial proximity enrichment or depletion of cell type pairs.

#### Value

ggplot heatmap

#### **Examples**

```
cellProximityHeatmap(CPscore)
```

```
cellProximityNetwork cellProximityNetwork
```

#### **Description**

Create network from cell-cell proximity scores

```
cellProximityNetwork(
  gobject,
  CPscore,
  remove_self_edges = FALSE,
  self_loop_strength = 0.1,
  color_depletion = "lightgreen",
  color_enrichment = "red",
  rescale_edge_weights = TRUE,
  edge_weight_range_depletion = c(0.1, 1),
  edge_weight_range_enrichment = c(1, 5),
  layout = "Fruchterman",
  only_show_enrichment_edges = F,
  edge_width_range = c(0.1, 2),
  node_size = 4,
  node_text_size = 6,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "cellProximityNetwork"
)
```

#### **Arguments**

gobject giotto object **CPscore** CPscore, output from cellProximityEnrichment() remove\_self\_edges remove enrichment/depletion edges with itself  $self_loop_strength$ size of self-loops color\_depletion color for depleted cell-cell interactions color\_enrichment color for enriched cell-cell interactions rescale\_edge\_weights rescale edge weights (boolean)  ${\tt edge\_weight\_range\_depletion}$ numerical vector of length 2 to rescale depleted edge weights edge\_weight\_range\_enrichment numerical vector of length 2 to rescale enriched edge weights layout layout algorithm to use to draw nodes and edges only\_show\_enrichment\_edges show only the enriched pairwise scores edge\_width\_range range of edge width size of nodes node\_size node\_text\_size size of node labels show\_plot show plot return\_plot return ggplot object save\_plot directly save the plot [boolean] save\_param list of saving parameters from all\_plots\_save\_function

## **Details**

This function creates a network that shows the spatial proximity enrichment or depletion of cell type pairs.

default save name for saving, don't change, change save\_name in save\_param

#### Value

igraph plot

default\_save\_name

# Examples

 ${\tt cellProximityNetwork(CPscore)}$ 

cellProximitySpatPlot 29

cellProximitySpatPlot cellProximitySpatPlot

## **Description**

Visualize 2D cell-cell interactions according to spatial coordinates in ggplot mode

## Usage

```
cellProximitySpatPlot(gobject, ...)
```

```
gobject
                  giotto object
interaction_name
                  cell-cell interaction name
cluster_column cluster column with cell clusters
sdimx
                  x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
show_other_cells
                  decide if show cells not in network
show_network
                  show underlying spatial network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
show_grid
                  show spatial grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
coord_fix_ratio
                  fix ratio between x and y-axis
show_legend
                  show legend
point_size_select
                  size of selected points
point_select_border_col
                  border color of selected points
point_select_border_stroke
                  stroke size of selected points
point_size_other
                  size of other points
```

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

cellProximitySpatPlot2D and cellProximitySpatPlot3D for 3D

### **Examples**

```
cellProximitySpatPlot(gobject)
```

```
cellProximitySpatPlot2D
```

cell Proximity Spat Plot 2D

# Description

Visualize 2D cell-cell interactions according to spatial coordinates in ggplot mode

```
cellProximitySpatPlot2D(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
```

```
spatial_network_name = "spatial_network",
  show_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  coord_fix_ratio = 1,
  show_legend = T,
 point_size_select = 2,
  point_select_border_col = "black",
  point_select_border_stroke = 0.05,
 point_size_other = 1,
 point_alpha_other = 0.3,
  point_other_border_col = "lightgrey",
 point_other_border_stroke = 0.01,
  show_plot = NA,
 return_plot = NA,
  save_plot = NA,
  save_param = list(),
 default_save_name = "cellProximitySpatPlot2D"
)
```

```
giotto object
gobject
interaction_name
                  cell-cell interaction name
cluster_column cluster column with cell clusters
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
                  color for cells (see details)
cell_color
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
show_other_cells
                  decide if show cells not in network
show_network
                  show underlying spatial network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
coord_fix_ratio
                  fix ratio between x and y-axis
show_legend
                  show legend
point_size_select
                  size of selected points
```

```
point_select_border_col
                 border color of selected points
point_select_border_stroke
                 stroke size of selected points
point_size_other
                 size of other points
point_other_border_col
                 border color of other points
point_other_border_stroke
                 stroke size of other points
show_plot
                 show plots
return_plot
                 return ggplot object
                 directly save the plot [boolean]
save_plot
save_param
                 list of saving parameters from all_plots_save_function
default_save_name
                 default save name for saving, don't change, change save_name in save_param
```

#### **Details**

Description of parameters.

#### Value

ggplot

# **Examples**

```
cellProximitySpatPlot2D(gobject)
```

```
cell Proximity Spat Plot 3D \\ cell Proximity Spat Plot 2D
```

## **Description**

Visualize 3D cell-cell interactions according to spatial coordinates in plotly mode

```
cellProximitySpatPlot3D(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = T,
```

```
show_network = T,
      show_other_network = F,
     network_color = NULL,
      spatial_network_name = "spatial_network",
      show_grid = F,
     grid_color = NULL,
      spatial_grid_name = "spatial_grid",
      show_legend = T,
      point_size_select = 4,
     point_size_other = 2,
     point_alpha_other = 0.5,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
     x_ticks = NULL,
     y_ticks = NULL,
      z_ticks = NULL,
      show_plot = NA,
     return_plot = NA,
      save_plot = NA,
     save_param = list(),
     default_save_name = "cellProximitySpatPlot3D",
   )
Arguments
   gobject
                    giotto object
   interaction_name
                    cell-cell interaction name
   cluster_column cluster column with cell clusters
   sdimx
```

```
x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
                  z-axis dimension name (default = 'sdimz')
sdimz
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
show_other_cells
                  decide if show cells not in network
                  show underlying spatial network
show_network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
show_grid
                  show spatial grid
                  color of spatial grid
grid_color
spatial_grid_name
                  name of spatial grid to use
show_legend
                  show legend
```

cellProximityVisPlot

```
point_size_select
size of selected points

point_size_other
size of other points

show_plot show plots

return_plot return plotly object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name
default save name for saving, don't change, change save_name in save_param
```

## **Details**

Description of parameters.

#### Value

plotly

## **Examples**

```
cellProximitySpatPlot3D(gobject)
```

```
cellProximityVisPlot cellProximityVisPlot
```

# **Description**

Visualize cell-cell interactions according to spatial coordinates

```
cellProximityVisPlot(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
```

cellProximityVisPlot 35

```
coord_fix_ratio = 1,
      show_legend = T,
      point_size_select = 2,
      point_select_border_col = "black",
      point_select_border_stroke = 0.05,
      point_size_other = 1,
      point_alpha_other = 0.3,
      point_other_border_col = "lightgrey",
      point_other_border_stroke = 0.01,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
      x_ticks = NULL,
      y_ticks = NULL,
      z_ticks = NULL,
      plot_method = c("ggplot", "plotly"),
    )
Arguments
   gobject
                     giotto object
    interaction_name
                     cell-cell interaction name
    cluster_column cluster column with cell clusters
                     x-axis dimension name (default = 'sdimx')
    sdimx
                     y-axis dimension name (default = 'sdimy')
    sdimy
    sdimz
                     z-axis dimension name (default = 'sdimz')
    cell_color
                     color for cells (see details)
    cell_color_code
                     named vector with colors
    color_as_factor
                     convert color column to factor
                     show underlying spatial network
    show_network
    network_color
                     color of spatial network
    spatial_network_name
                     name of spatial network to use
    show_grid
                     show spatial grid
    grid_color
                     color of spatial grid
    spatial_grid_name
                     name of spatial grid to use
    coord_fix_ratio
                     fix ratio between x and y-axis
    show\_legend
                     show legend
   point_size_select
                     size of selected points
   {\tt point\_select\_border\_col}
```

border color of selected points

## **Details**

Description of parameters.

#### Value

ggplot or plotly

#### **Examples**

```
cellProximityVisPlot(gobject)
```

```
cell Proximity VisPlot\_2D\_ggplot \\ cell Proximity VisPlot\_2D\_ggplot
```

## **Description**

Visualize 2D cell-cell interactions according to spatial coordinates in ggplot mode

```
cellProximityVisPlot_2D_ggplot(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = NULL,
  sdimy = NULL,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  show_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  coord_fix_ratio = 1,
  show_legend = T,
  point_size_select = 2,
```

```
point_select_border_col = "black",
 point_select_border_stroke = 0.05,
 point_size_other = 1,
 point_alpha_other = 0.3,
 point_other_border_col = "lightgrey",
 point_other_border_stroke = 0.01,
)
```

#### **Arguments**

```
gobject
                  giotto object
interaction_name
                  cell-cell interaction name
cluster_column cluster column with cell clusters
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
show_other_cells
                  decide if show cells not in network
show_network
                  show underlying spatial network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
coord_fix_ratio
                  fix ratio between x and y-axis
show_legend
                  show legend
point_size_select
                  size of selected points
point_select_border_col
                  border color of selected points
point_select_border_stroke
                  stroke size of selected points
point_size_other
                  size of other points
point_other_border_col
                  border color of other points
\verb"point_other_border_stroke"
                  stroke size of other points
```

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

```
cellProximityVisPlot_2D_ggplot(gobject)
```

```
cell Proximity VisPlot\_2D\_plotly \\ cell Proximity VisPlot\_2D\_plotly
```

## Description

Visualize 2D cell-cell interactions according to spatial coordinates in plotly mode

```
cellProximityVisPlot_2D_plotly(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = NULL,
  sdimy = NULL,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  show_legend = T,
  point_size_select = 2,
  point_size_other = 1,
  point_alpha_other = 0.3,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_{ticks} = NULL,
 y_ticks = NULL,
)
```

## **Arguments**

gobject giotto object

interaction\_name

cell-cell interaction name

cluster\_column cluster column with cell clusters

sdimx x-axis dimension name (default = 'sdimx')

sdimy y-axis dimension name (default = 'sdimy')

cell\_color color for cells (see details)

cell\_color\_code

named vector with colors

color\_as\_factor

convert color column to factor

show\_other\_cells

decide if show cells not in network

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

show\_grid show spatial grid

grid\_color color of spatial grid

spatial\_grid\_name

name of spatial grid to use

show\_legend show legend

point\_size\_select

size of selected points

coord\_fix\_ratio

fix ratio between x and y-axis

## **Details**

Description of parameters.

#### Value

plotly

## **Examples**

cellProximityVisPlot\_2D\_plotly(gobject)

```
cell Proximity VisPlot\_3D\_plotly \\ cell Proximity VisPlot\_3D\_plotly
```

## **Description**

Visualize 3D cell-cell interactions according to spatial coordinates in plotly mode

## Usage

```
cellProximityVisPlot_3D_plotly(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  show_legend = T,
  point_size_select = 2,
  point_size_other = 1,
  point_alpha_other = 0.5,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_{ticks} = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
)
```

## **Arguments**

```
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
show_other_cells
                  decide if show cells not in network
show_network
                  show underlying spatial network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
show_grid
                  show spatial grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
                  show legend
show_legend
point_size_select
                  size of selected points
coord_fix_ratio
                  fix ratio between x and y-axis
```

## **Details**

Description of parameters.

## Value

plotly

## **Examples**

```
cellProximityVisPlot_3D_plotly(gobject)
```

 ${\tt change} {\tt GiottoInstructions}$ 

change Giot to Instructions

## Description

Function to change one or more instructions from giotto object

```
changeGiottoInstructions(
  gobject,
  params = NULL,
  new_values = NULL,
  return_gobject = TRUE
)
```

42 clusterCells

#### **Arguments**

```
gobject giotto object
params parameter(s) to change
new_values new value(s) for parameter(s)
return_gobject (boolean) return giotto object
```

#### Value

named vector with giotto instructions

## **Examples**

changeGiottoInstructions()

clusterCells

clusterCells

#### **Description**

cluster cells using a variety of different methods

```
clusterCells(
  gobject,
 cluster_method = c("leiden", "louvain_community", "louvain_multinet", "randomwalk",
    "sNNclust", "kmeans", "hierarchical"),
  name = "cluster_name",
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  pyth_leid_resolution = 1,
  pyth_leid_weight_col = "weight",
 pyth_leid_part_type = c("RBConfigurationVertexPartition", "ModularityVertexPartition"),
  pyth_leid_init_memb = NULL,
  pyth_leid_iterations = 1000,
  pyth_louv_resolution = 1,
  pyth_louv_weight_col = NULL,
  python_louv_random = F,
  python_path = NULL,
  louvain_gamma = 1,
  louvain_omega = 1,
  walk\_steps = 4,
  walk_clusters = 10,
  walk_weights = NA,
  sNNclust_k = 20,
  sNNclust_eps = 4,
  sNNclust_minPts = 16,
  borderPoints = TRUE,
  expression_values = c("normalized", "scaled", "custom"),
  genes_to_use = NULL,
```

clusterCells 43

```
dim_reduction_to_use = c("cells", "pca", "umap", "tsne"),
      dim_reduction_name = "pca",
      dimensions_to_use = 1:10,
      distance_method = c("original", "pearson", "spearman", "euclidean", "maximum",
        "manhattan", "canberra", "binary", "minkowski"),
      km_centers = 10,
      km_iter_max = 100,
      km_nstart = 1000,
      km_algorithm = "Hartigan-Wong",
     hc_agglomeration_method = c("ward.D2", "ward.D", "single", "complete", "average",
        "mcquitty", "median", "centroid"),
      hc_k = 10,
      hc_h = NULL
      return_gobject = TRUE,
      set_seed = T,
      seed_number = 1234
Arguments
   gobject
                    giotto object
    cluster_method community cluster method to use
                    name for new clustering result
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
    network_name
                    name of NN network to use
   pyth_leid_resolution
                    resolution for leiden
    pyth_leid_weight_col
                    column to use for weights
   pyth_leid_part_type
                    partition type to use
   pyth_leid_init_memb
                    initial membership
    pyth_leid_iterations
                    number of iterations
    pyth_louv_resolution
                    resolution for louvain
   pyth_louv_weight_col
                    python louvain param: weight column
    python_louv_random
                    python louvain param: random
    python_path
                    specify specific path to python if required
    louvain_gamma
                    louvain param: gamma or resolution
                    louvain param: omega
    louvain_omega
                    randomwalk: number of steps
    walk_steps
                    randomwalk: number of clusters
   walk_clusters
                    randomwalk: weight column
    walk_weights
    sNNclust_k
                    SNNclust: k neighbors to use
```

44 clusterCells

sNNclust\_eps SNNclust: epsilon

sNNclust\_minPts

SNNclust: min points

borderPoints SNNclust: border points

expression\_values

expression values to use

genes\_to\_use = NULL,
dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

name of reduction 'pca',

dimensions\_to\_use

dimensions to use

distance\_method

distance method

km\_centers kmeans centers km\_iter\_max kmeans iterations

km\_nstart kmeans random starting points

 ${\tt km\_algorithm} \quad \ \, {\tt kmeans} \; {\tt algorithm} \\$ 

hc\_agglomeration\_method

hierarchical clustering method

hc\_k hierachical number of clusters

hc\_h hierarchical tree cutoff

 $\verb|return_gobject|| boolean: return giotto object (default = TRUE)$ 

set\_seed set seed

seed\_number number for seed

## **Details**

Wrapper for the different clustering methods.

## Value

giotto object with new clusters appended to cell metadata

#### See Also

 $\label{lem:cluster_doLouvainCluster_multinet} do Louvain Cluster\_community, do Louvain Cluster\_multinet, do Louvain Cluster\_do Random Walk Cluster\_do SNN Cluster\_do Kmeans\_do H clust$ 

## **Examples**

clusterCells(gobject)

clusterSpatialCorGenes

```
clusterSpatialCorGenes
```

clusterSpatialCorGenes

## **Description**

Cluster based on spatially correlated genes

## Usage

```
clusterSpatialCorGenes(
   spatCorObject,
   name = "spat_clus",
   hclust_method = "ward.D",
   k = 10,
   return_obj = TRUE
)
```

## Arguments

```
spatCorObject spatial correlation object
```

name name for spatial clustering results
hclust\_method method for hierarchical clustering
k number of clusters to extract

return\_obj return spatial correlation object (spatCorObject)

#### Value

spatCorObject or cluster results

#### **Examples**

```
clusterSpatialCorGenes(gobject)
```

combineMetadata

combineMetadata

## Description

This function combines the cell metadata with spatial locations and enrichment results from createSpatialEnrich

## Usage

```
combineMetadata(gobject, spat_enr_names = NULL)
```

## **Arguments**

```
gobject Giotto object
spat_enr_names names of spatial enrichment results to include
```

46 createGiottoInstructions

#### Value

Extended cell metadata in data.table format.

## **Examples**

```
combineMetadata(gobject)
```

convertEnsemblToGeneSymbol

convert Ensembl To Gene Symbol

## Description

This function convert ensembl gene IDs from a matrix to official gene symbols

## Usage

```
convertEnsemblToGeneSymbol(matrix, species = c("mouse", "human"))
```

## **Arguments**

matrix an expression matrix with ensembl gene IDs as rownames

species species to use for gene symbol conversion

## **Details**

This function requires that the biomaRt library is installed

#### Value

expression matrix with gene symbols as rownames

# Examples

```
convertEnsemblToGeneSymbol(matrix)
```

create Giotto Instructions

createGiottoInstructions

## Description

Function to set global instructions for giotto functions

createGiottoObject 47

#### Usage

```
createGiottoInstructions(
   python_path = NULL,
   show_plot = NULL,
   return_plot = NULL,
   save_plot = NULL,
   save_dir = NULL,
   plot_format = NULL,
   dpi = NULL,
   units = NULL,
   height = NULL,
   width = NULL
```

#### **Arguments**

path to python binary to use python\_path show\_plot print plot to console, default = TRUE return plot as object, default = TRUE return\_plot save\_plot automatically save plot, dafault = FALSE path to directory where to save plots save\_dir dpi resolution for raster images height height of plots width width of plots

#### Value

named vector with giotto instructions

## **Examples**

```
createGiottoInstructions()
```

## Description

Function to create a giotto object

```
createGiottoObject(
  raw_exprs,
  spatial_locs = NULL,
  norm_expr = NULL,
  norm_scaled_expr = NULL,
  custom_expr = NULL,
  cell_metadata = NULL,
```

48 createGiottoObject

```
gene_metadata = NULL,
spatial_network = NULL,
spatial_network_name = NULL,
spatial_grid = NULL,
spatial_grid_name = NULL,
spatial_enrichment = NULL,
spatial_enrichment_name = NULL,
dimension_reduction = NULL,
nn_network = NULL,
offset_file = NULL,
instructions = NULL
```

#### Arguments

```
raw_exprs
                  matrix with raw expression counts [required]
                  data.table or data.frame with coordinates for cell centroids
spatial_locs
                  normalized expression values
norm_expr
norm_scaled_expr
                  scaled expression values
custom_expr
                  custom expression values
cell_metadata
                  cell annotation metadata
gene_metadata
                  gene annotation metadata
spatial_network
                  list of spatial network(s)
spatial_network_name
                  list of spatial network name(s)
                  list of spatial grid(s)
spatial_grid
spatial_grid_name
                  list of spatial grid name(s)
spatial_enrichment
                  list of spatial enrichment score(s) for each spatial region
spatial_enrichment_name
                  list of spatial enrichment name(s)
dimension_reduction
                  list of dimension reduction(s)
                  list of nearest neighbor network(s)
nn_network
                  file used to stitch fields together (optional)
offset_file
instructions
                  list of instructions or output result from createGiottoInstructions
```

#### **Details**

[Requirements] To create a giotto object you need to provide at least a matrix with genes as row names and cells as column names. To include spatial information about cells (or regions) you need to provide a data.table or data.frame with coordinates for all spatial dimensions. This can be 2D (x and y) or 3D (x, y, x). The row order for the cell coordinates should be the same as the column order for the provided expression data.

[Instructions] Additionally an instruction file, generated manually or with createGiottoInstructions can be provided to instructions, if not a default instruction file will be created for the Giotto object.

createHeatmap\_DT 49

[Multiple fields] In case a dataset consists of multiple fields, like seqFISH+ for example, an offset file can be provided to stitch the different fields together. stitchFieldCoordinates can be used to generate such an offset file.

[**Processed data**] Processed count data, such as normalized data, can be provided using one of the different expression slots (norm\_expr, norm\_scaled\_expr, custom\_expr).

[Metadata] Cell and gene metadata can be provided using the cell and gene metadata slots. This data can also be added afterwards using the addGeneMetadata or addCellMetadata functions.

[Other information] Additional information can be provided through the appropriate slots:

- · spatial networks
- · spatial girds
- · spatial enrichments
- · dimensions reductions
- · nearest neighbours networks

#### Value

giotto object

## **Examples**

```
createGiottoObject(raw_exprs, spatial_locs)
```

createHeatmap\_DT

createHeatmap\_DT

#### **Description**

creates order for clusters

```
createHeatmap_DT(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  cluster_column = NULL,
  cluster_order = c("size", "correlation", "custom"),
  cluster_custom_order = NULL,
  cluster_cor_method = "pearson",
  cluster_hclust_method = "ward.D",
  gene_order = c("custom", "correlation"),
  gene_custom_order = NULL,
  gene_cor_method = "pearson",
  gene_hclust_method = "complete"
)
```

50 createMetagenes

## **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
genes
                  genes to use
cluster_column name of column to use for clusters
cluster_order method to determine cluster order
cluster_custom_order
                  custom order for clusters
cluster_cor_method
                  method for cluster correlation
{\tt cluster\_hclust\_method}
                  method for hierarchical clustering of clusters
gene_order
                  method to determine gene order
gene_custom_order
                  custom order for genes
gene_cor_method
                  method for gene correlation
gene_hclust_method
                  method for hierarchical clustering of genes
```

#### **Details**

Creates input data.tables for plotHeatmap function.

#### Value

list

## **Examples**

```
createHeatmap_DT(gobject)
```

createMetagenes

createMetagenes

## Description

This function creates an average metagene for gene clusters.

```
createMetagenes(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  gene_clusters,
  name = "metagene",
  return_gobject = TRUE
)
```

createNearestNetwork 51

## **Arguments**

## **Details**

```
An example for the 'gene_clusters' could be like this: cluster_vector = c(1, 1, 2, 2); names(cluster_vector) = c('geneA', 'geneB', 'geneC', 'geneD')
```

## Value

giotto object

#### **Examples**

```
createMetagenes(gobject)
```

createNearestNetwork createNearestNetwork

# Description

create a nearest neighbour (NN) network

```
createNearestNetwork(
  gobject,
  type = c("sNN", "kNN"),
  dim_reduction_to_use = "pca",
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  genes_to_use = NULL,
  expression_values = c("normalized", "scaled", "custom"),
  name = "sNN.pca",
  return_gobject = TRUE,
  k = 30,
  minimum_shared = 5,
  top_shared = 3,
  verbose = T,
  ...
)
```

52 createNearestNetwork

#### **Arguments**

gobject giotto object type sNN or kNN

dim\_reduction\_to\_use

dimension reduction method to use

dim\_reduction\_name

name of dimension reduction set to use

dimensions\_to\_use

number of dimensions to use as input

genes\_to\_use if dim\_reduction\_to\_use = NULL, which genes to use

expression\_values

expression values to use

name arbitrary name for NN network

return\_gobject boolean: return giotto object (default = TRUE)

k number of k neighbors to use minimum\_shared minimum shared neighbors

top\_shared keep at ...
verbose be verbose

additional parameters for kNN and sNN functions from dbscan

#### **Details**

This function creates a k-nearest neighbour (kNN) or shared nearest neighbour (sNN) network based on the provided dimension reduction space. To run it directly on the gene expression matrix set  $dim\_reduction\_to\_use = NULL$ .

See also kNN and sNN for more information about how the networks are created.

Output for kNN:

• from: cell ID for source cell

• to: cell\_ID for target cell

• distance: distance between cells

• weight: weight = 1/(1 + distance)

#### Output for sNN:

• from: cell\_ID for source cell

• to: cell\_ID for target cell

• distance: distance between cells

• weight: 1/(1 + distance)

• shared: number of shared neighbours

• rank: ranking of pairwise cell neighbours

For sNN networks two additional parameters can be set:

- minimum\_shared: minimum number of shared neighbours needed
- top\_shared: keep this number of the top shared neighbours, irrespective of minimum\_shared setting

createSpatialEnrich 53

#### Value

giotto object with updated NN network

#### **Examples**

```
createNearestNetwork(gobject)
```

createSpatialEnrich createSpatialEnrich

## **Description**

Function to calculate gene signature enrichment scores per spatial position using a hypergeometric test.

# Usage

```
createSpatialEnrich(
  gobject,
  enrich_method = c("PAGE", "rank", "hypergeometric"),
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  output_enrichment = c("original", "zscore"),
  name = "PAGE",
  return_gobject = TRUE
)
```

## Arguments

```
gobject
                  Giotto object
enrich_method method for gene signature enrichment calculation
                  Matrix of signature genes for each cell type / process
sign_matrix
expression_values
                  expression values to use
reverse_log_scale
                  reverse expression values from log scale
                  log base to use if reverse_log_scale = TRUE
logbase
output_enrichment
                  how to return enrichment output
                  to give to spatial enrichment results, default = PAGE
name
return_gobject return giotto object
```

#### **Details**

For details see the individual functions:

PAGE: PAGEEnrichPAGE: rankEnrich

• PAGE: hyperGeometricEnrich

54 createSpatialGrid

#### Value

Giotto object or enrichment results if return\_gobject = FALSE

## **Examples**

```
createSpatialEnrich(gobject)
```

createSpatialGrid

createSpatialGrid

#### **Description**

Create a spatial grid.

## Usage

```
createSpatialGrid(
  gobject,
  sdimx_stepsize = NULL,
  sdimy_stepsize = NULL,
  sdimz_stepsize = NULL,
  minimum_padding = 1,
  name = "spatial_grid",
  return_gobject = TRUE
)
```

## Arguments

## **Details**

Creates a spatial grid with defined x, y (and z) dimensions. The dimension units are based on the provided spatial location units.

#### Value

giotto object with updated spatial grid slot

```
{\tt createSpatialGrid(gobject)}
```

createSpatialGrid\_2D 55

```
createSpatialGrid\_2D createSpatialGrid\_2D
```

## Description

create a spatial grid for 2D spatial data.

# Usage

```
createSpatialGrid_2D(
  gobject,
  sdimx_stepsize = NULL,
  sdimy_stepsize = NULL,
  minimum_padding = 1,
  name = "spatial_grid",
  return_gobject = TRUE
)
```

## **Arguments**

## **Details**

Creates a spatial grid with defined x, y (and z) dimensions. The dimension units are based on the provided spatial location units.

## Value

giotto object with updated spatial grid slot

```
createSpatialGrid_2D(gobject)
```

```
createSpatialGrid\_3D \quad \textit{createSpatialGrid\_3D}
```

## **Description**

Create a spatial grid for 3D spatial data.

## Usage

```
createSpatialGrid_3D(
  gobject,
  sdimx_stepsize = NULL,
  sdimy_stepsize = NULL,
  sdimz_stepsize = NULL,
  minimum_padding = 1,
  name = "spatial_grid",
  return_gobject = TRUE
)
```

## **Arguments**

```
gobject giotto object

sdimx_stepsize stepsize along the x-axis

sdimy_stepsize stepsize along the y-axis

sdimz_stepsize stepsize along the z-axis

minimum_padding

minimum padding on the edges

name name for spatial grid (default = 'spatial_grid')

return_gobject boolean: return giotto object (default = TRUE)
```

## **Details**

Creates a spatial grid with defined x, y (and z) dimensions. The dimension units are based on the provided spatial location units.

#### Value

giotto object with updated spatial grid slot

```
createSpatialGrid_3D(gobject)
```

createSpatialNetwork 57

```
createSpatialNetwork createSpatialNetwork
```

#### **Description**

Create a spatial network based on cell centroid physical distances.

## Usage

```
createSpatialNetwork(
  gobject,
  k = 4,
  dimensions = "all",
  maximum_distance = NULL,
  minimum_k = 0,
  name = "spatial_network",
  verbose = F,
  return_gobject = TRUE
)
```

#### **Arguments**

```
gobject giotto object
k number of nearest neighbors based on physical distance
dimensions which spatial dimensions to use (default = all)
maximum_distance
distance cuttof for nearest neighbors to consider
minimum_k minimum nearest neighbours if maximum_distance != NULL
name name for spatial network (default = 'spatial_network')
verbose verbose
```

return\_gobject boolean: return giotto object (default = TRUE)

#### **Details**

Creates a spatial network connecting single-cells based on their physical distance to each other. Number of neighbors can be determined by k, maximum distance from each cell with or without setting a minimum k for each cell.

**dimensions:** default = 'all' which takes all possible dimensions. Alternatively you can provide a character vector that specififies the spatial dimensions to use, e.g. c("sdimx', "sdimy") or a numerical vector, e.g. 2:3

**maximum\_distance:** to create a network based on maximum distance only, you also need to set k to a very high value, e.g. k = 100

#### Value

giotto object with updated spatial network slot

```
createSpatialNetwork(gobject)
```

58 create\_average\_DT

```
create\_average\_detection\_DT \\ create\_average\_detection\_DT
```

## **Description**

calculates average gene detection for a cell metadata factor (e.g. cluster)

## Usage

```
create_average_detection_DT(
  gobject,
  meta_data_name,
  expression_values = c("normalized", "scaled", "custom"),
  detection_threshold = 0
)
```

#### **Arguments**

```
gobject giotto object

meta_data_name name of metadata column to use
expression_values

which expression values to use

detection_threshold

detection threshold to consider a gene detected
```

## Value

data.table with average gene epression values for each factor

## Description

calculates average gene expression for a cell metadata factor (e.g. cluster)

## Usage

```
create_average_DT(
  gobject,
  meta_data_name,
  expression_values = c("normalized", "scaled", "custom")
)
```

# Arguments

```
gobject giotto object
meta_data_name name of metadata column to use
expression_values
    which expression values to use
```

#### Value

data.table with average gene epression values for each factor

## Description

creates randomized cell ids within a selection of cell types

## Usage

```
create_cell_type_random_cell_IDs(
  gobject,
  cluster_column = "cell_types",
  needed_cell_types
)
```

## **Arguments**

```
gobject giotto object to use

cluster_column cluster column with cell type information

needed_cell_types

vector of cell type names for which a random id will be found
```

## **Details**

Details will follow.

## Value

list of randomly sampled cell ids with same cell type composition

```
create_cell_type_random_cell_IDs(gobject)
```

60 create\_dimObject

```
create_cluster_matrix create_cluster_matrix
```

## Description

creates aggregated matrix for a given clustering

## Usage

```
create_cluster_matrix(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  gene_subset = NULL
)
```

## **Examples**

```
create_cluster_matrix(gobject)
```

create\_dimObject

create\_dimObject

## Description

Creates an object that stores a dimension reduction output

## Usage

```
create_dimObject(
  name = "test",
  reduction_method = NULL,
  coordinates = NULL,
  misc = NULL,
  my_rownames = NULL
)
```

## Arguments

```
name arbitrary name for object
reduction_method
method used to reduce dimensions
coordinates accepts the coordinates after dimension reduction
misc any additional information will be added to this slot
```

## Value

number of distinct colors

decide\_cluster\_order 61

```
decide_cluster_order
```

## **Description**

creates order for clusters

## Usage

```
decide_cluster_order(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  cluster_column = NULL,
  cluster_order = c("size", "correlation", "custom"),
  cluster_custom_order = NULL,
  cor_method = "pearson",
  hclust_method = "ward.D"
)
```

## **Arguments**

```
gobject giotto object
expression_values
expression values to use
genes genes to use
cluster_column name of column to use for clusters
cluster_order method to determine cluster order
cluster_custom_order
custom order for clusters

cor_method method for correlation
hclust_method method for hierarchical clustering
```

## **Details**

Calculates order for clusters.

#### Value

custom

```
decide_cluster_order(gobject)
```

```
{\tt detectSpatialCorGenes} \ \ \textit{detectSpatialCorGenes}
```

#### **Description**

Detect genes that are spatially correlated

## Usage

```
detectSpatialCorGenes(
  gobject,
  method = c("grid", "network"),
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_network_name = "spatial_network",
  network_smoothing = NULL,
  spatial_grid_name = "spatial_grid",
  min_cells_per_grid = 4,
  cor_method = c("pearson", "kendall", "spearman")
)
```

#### **Arguments**

```
giotto object
gobject
                  method to use for spatial averaging
method
expression_values
                  gene expression values to use
subset_genes
                  subset of genes to use
spatial_network_name
                  name of spatial network to use
network_smoothing
                  smoothing factor beteen 0 and 1 (default: automatic)
spatial_grid_name
                  name of spatial grid to use
min_cells_per_grid
                  minimum number of cells to consider a grid
b
                  smoothing factor beteen 0 and 1 (default: automatic)
```

#### **Details**

- 1. grid-averaging: average gene expression values within a predefined spatial grid
- 2. network-averaging: smoothens the gene expression matrix by averaging the expression within one cell by using the neighbours within the predefined spatial network. b is a smoothening factor that defaults to 1 1/k, where k is the median number of k-neighbors in the selected spatial network. Setting b = 0 means no smoothing and b = 1 means no contribution from its own expression.

The spatCorObject can be further explored with showSpatialCorGenes()

detectSpatialPatterns 63

#### Value

```
returns a spatial correlation object: "spatCorObject"
```

#### See Also

```
showSpatialCorGenes
```

## **Examples**

```
detectSpatialCorGenes(gobject)
```

```
detectSpatialPatterns detectSpatialPatterns
```

## **Description**

Identify spatial patterns through PCA on average expression in a spatial grid.

# Usage

```
detectSpatialPatterns(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  spatial_grid_name = "spatial_grid",
  min_cells_per_grid = 4,
  scale_unit = F,
  ncp = 100,
  show_plot = T,
  PC_zscore = 1.5
)
```

## Arguments

```
gobject
                  giotto object
expression_values
                  expression values to use
spatial_grid_name
                  name of spatial grid to use (default = 'spatial_grid')
min_cells_per_grid
                  minimum number of cells in a grid to be considered
                  scale features
scale_unit
                  number of principal components to calculate
ncp
show_plot
                  show plots
PC_zscore
                  minimum z-score of variance explained by a PC
```

64 dimCellPlot

#### **Details**

Steps to identify spatial patterns:

- 1. average gene expression for cells within a grid, see createSpatialGrid
- 2. perform PCA on the average grid expression profiles
- 3. convert variance of principlal components (PCs) to z-scores and select PCs based on a z-score threshold

#### Value

```
spatial pattern object 'spatPatObj'
```

## **Examples**

```
detectSpatialPatterns(gobject)
```

dimCellPlot

dimCellPlot

#### **Description**

Visualize cells according to dimension reduction coordinates

```
dimCellPlot(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  cell_annotation_values,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
```

dimCellPlot 65

```
edge_alpha = NULL,
      point_size = 1,
      point_border_col = "black",
      point_border_stroke = 0.1,
      show_legend = T,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimCellPlot"
Arguments
    gobject
                     giotto object
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
    dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    show_other_cells
                     display not selected cells
    other_cell_color
```

color of not selected cells

dimCellPlot

```
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
                  column to use for alpha of the edges
edge_alpha
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
                  show plot
show_plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
title
                  title for plot, defaults to cell_color parameter
```

# Details

Description of parameters. For 3D plots see dimCellPlot2D

#### Value

ggplot

```
dimCellPlot(gobject)
```

dimCellPlot2D 67

dimCellPlot2D

dimCellPlot2D

#### **Description**

Visualize cells according to dimension reduction coordinates

```
dimCellPlot2D(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  cell_annotation_values,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show\_center\_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  legend_text = 8,
  axis_text = 8,
  axis_title = 8,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
```

68 dimCellPlot2D

```
default_save_name = "dimCellPlot2D"
Arguments
    gobject
                     giotto object
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    show_other_cells
                     display not selected cells
    other_cell_color
                     color of not selected cells
    other_point_size
                     size of not selected cells
    show_cluster_center
                     plot center of selected clusters
    show_center_label
                     plot label of selected clusters
    center_point_size
                     size of center points
    label_size
                     size of labels
    label_fontface font of labels
                     column to use for alpha of the edges
    edge_alpha
                     size of point (cell)
    point_size
```

color of border around points

point\_border\_col

dimGenePlot 69

```
point_border_stroke
                  stroke size of border around points
                  show legend
show_legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
title
                  title for plot, defaults to cell_color parameter
```

#### **Details**

Description of parameters. For 3D plots see dimPlot3D

#### Value

ggplot

## **Examples**

```
dimCellPlot2D(gobject)
```

dimGenePlot

dimGenePlot

## **Description**

Visualize cells and gene expression according to dimension reduction coordinates

```
dimGenePlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  show_NN_network = F,
```

70 dimGenePlot

```
nn_network_to_use = "sNN",
      network_name = "sNN.pca",
      network_color = "lightgray",
      edge_alpha = NULL,
      scale_alpha_with_expression = FALSE,
      point_size = 1,
      genes_high_color = "red",
      genes_mid_color = "white",
      genes_low_color = "blue",
      point_border_col = "black",
      point_border_stroke = 0.1,
      midpoint = 0,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h"
      show_legend = T,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot"
    )
Arguments
   gobject
                    giotto object
    expression_values
                    gene expression values to use
                    genes to show
    genes
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
                    dimension to use on x-axis
    dim1_to_use
   dim2_to_use
                    dimension to use on y-axis
    show_NN_network
                    show underlying NN network
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
                    name of NN network to use, if show_NN_network = TRUE
   network_name
    edge_alpha
                    column to use for alpha of the edges
    scale_alpha_with_expression
                    scale expression with ggplot alpha parameter
                    size of point (cell)
   point_size
    point_border_col
                    color of border around points
   point_border_stroke
```

stroke size of border around points

dimGenePlot2D 71

```
midpoint
                  size of point (cell)
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_legend
                  show legend
show_plot
                  show plots
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  parameters for cowplot::save_plot()
```

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

```
dimGenePlot3D
```

## **Examples**

```
dimGenePlot(gobject)
```

dimGenePlot2D

dimGenePlot2D

## Description

Visualize cells and gene expression according to dimension reduction coordinates

```
dimGenePlot2D(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  show_NN_network = F,
  nn_network_to_use = "sNN",
```

72 dimGenePlot2D

```
network_name = "sNN.pca",
      network_color = "lightgray",
      edge_alpha = NULL,
      scale_alpha_with_expression = FALSE,
      point_size = 1,
      genes_high_color = "red",
      genes_mid_color = "white",
      genes_low_color = "blue",
      point_border_col = "black",
      point_border_stroke = 0.1,
      midpoint = 0,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_legend = T,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot2D"
    )
Arguments
   gobject
                    giotto object
    expression_values
                    gene expression values to use
    genes
                    genes to show
    dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
    dim1_to_use
                    dimension to use on x-axis
    dim2_to_use
                    dimension to use on y-axis
    show_NN_network
                    show underlying NN network
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
                    name of NN network to use, if show_NN_network = TRUE
   network_name
    edge_alpha
                    column to use for alpha of the edges
    scale_alpha_with_expression
                    scale expression with ggplot alpha parameter
    point_size
                    size of point (cell)
   point_border_col
                    color of border around points
   point_border_stroke
                    stroke size of border around points
   midpoint
                    size of point (cell)
```

dimGenePlot3D 73

```
cowplot param: how many columns
cow_n_col
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
                 cowplot param: how to align
cow_align
                 show legend
show_legend
show_plot
                 show plots
return_plot
                 return ggplot object
save_plot
                 directly save the plot [boolean]
                 list of saving parameters from all_plots_save_function
save_param
default_save_name
                 default save name for saving, don't change, change save_name in save_param
                 parameters for cowplot::save_plot()
```

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

dimGenePlot3D

### **Examples**

```
dimGenePlot2D(gobject)
```

dimGenePlot3D

dimGenePlot3D

### **Description**

Visualize cells and gene expression according to dimension reduction coordinates

```
dimGenePlot3D(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  dim3_to_use = 3,
  show_NN_network = F,
  nn_network_to_use = "sNN",
```

74 dimGenePlot3D

```
network_name = "sNN.pca",
      network_color = "lightgray",
      cluster_column = NULL,
      select_cell_groups = NULL,
      select_cells = NULL,
      show_other_cells = T,
      other_cell_color = "lightgrey",
      other_point_size = 1,
      edge_alpha = NULL,
      point_size = 2,
      genes_high_color = NULL,
      genes_mid_color = "white",
      genes_low_color = "blue",
      show_legend = T,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot3D"
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
    genes
                     genes to show
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
                     dimension to use on y-axis
    dim2_to_use
                     dimension to use on z-axis
    dim3_to_use
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
   network_name
    edge_alpha
                     column to use for alpha of the edges
                     size of point (cell)
   point_size
    show_legend
                     show legend
    show_plot
                     show plots
                     return ggplot object
    return_plot
    save_plot
                     directly save the plot [boolean]
                     list of saving parameters from all_plots_save_function
    save_param
    default_save_name
                     default save name for saving, don't change, change save_name in save_param
                     parameters for cowplot::save_plot()
```

. . .

dimPlot 75

#### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

dimGenePlot3D(gobject)

dimPlot

dimPlot

### **Description**

Visualize cells according to dimension reduction coordinates

```
dimPlot(
  gobject,
  group_by = NULL,
  group_by_subset = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
```

76 dimPlot

```
point_size = 1,
      point_border_col = "black",
      point_border_stroke = 0.1,
      show_legend = T,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      title = NULL,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimPlot"
    )
Arguments
    gobject
                     giotto object
    group_by_subset
                     subset the group_by factor column
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
    dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
```

dimPlot 77

```
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
                  size of legend text
legend_text
axis_text
                  size of axis text
axis_title
                  size of axis title
title
                  title for plot, defaults to cell_color parameter
cow_n_col
                  cowplot param: how many columns
                  cowplot param: relative height
cow_rel_h
cow_rel_w
                  cowplot param: relative width
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  create multiple plots based on cell annotation column
groub_by
```

## Details

Description of parameters, see dimPlot2D. For 3D plots see dimPlot3D

#### Value

ggplot

#### **Examples**

```
dimPlot(gobject)
```

78 dimPlot2D

dimPlot2D

dimPlot2D

#### **Description**

Visualize cells according to dimension reduction coordinates

```
dimPlot2D(
  gobject,
  group_by = NULL,
  group_by_subset = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  title = NULL,
  show_legend = T,
  legend_text = 8,
  axis_text = 8,
  axis_title = 8,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
```

dimPlot2D 79

```
cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimPlot2D"
Arguments
    gobject
                     giotto object
    group_by_subset
                     subset the group_by factor column
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
                     dimension to use on y-axis
    dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    show_other_cells
                     display not selected cells
    other_cell_color
                     color of not selected cells
    other_point_size
                     size of not selected cells
    show_cluster_center
                     plot center of selected clusters
    show_center_label
                     plot label of selected clusters
```

80 dimPlot2D

center\_point\_size

size of center points

label\_size size of labels label\_fontface font of labels

edge\_alpha column to use for alpha of the edges

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

title title for plot, defaults to cell\_color parameter

show\_legend show legend

legend\_text size of legend text
axis\_text size of axis text
axis\_title size of axis title

cow\_n\_col cowplot param: how many columns

cow\_rel\_h cowplot param: relative height
cow\_rel\_w cowplot param: relative width
cow\_align cowplot param: how to align

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

groub\_by create multiple plots based on cell annotation column

### **Details**

Description of parameters. For 3D plots see dimPlot3D

### Value

ggplot

### **Examples**

dimPlot2D(gobject)

dimPlot2D\_single 81

dimPlot2D\_single

dimPlot2D\_single

### **Description**

Visualize cells according to dimension reduction coordinates

```
dimPlot2D_single(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  title = NULL,
  show_legend = T,
  legend_text = 8,
  axis_text = 8,
  axis_title = 8,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "dimPlot2D_single"
```

82 dimPlot2D\_single

)

### Arguments

giotto object gobject dim\_reduction\_to\_use dimension reduction to use dim\_reduction\_name dimension reduction name dim1\_to\_use dimension to use on x-axis dim2\_to\_use dimension to use on y-axis spat\_enr\_names names of spatial enrichment results to include show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) network\_name name of NN network to use, if show\_NN\_network = TRUE cell\_color color for cells (see details) color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_size size of point (cell)

dimPlot3D 83

```
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
title
                  title for plot, defaults to cell_color parameter
show_legend
                  show legend
legend\_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

#### **Details**

Description of parameters. For 3D plots see dimPlot3D

#### Value

ggplot

### **Examples**

```
dimPlot2D_single(gobject)
```

dimPlot3D

dimPlot3D

### **Description**

Visualize cells according to dimension reduction coordinates

```
dimPlot3D(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  dim3_to_use = 3,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 2,
```

84 dimPlot3D

```
show_NN_network = F,
      nn_network_to_use = "sNN",
      network_name = "sNN.pca",
      color_as_factor = T,
      cell_color = NULL,
      cell_color_code = NULL,
      show_cluster_center = F,
      show_center_label = T,
      center_point_size = 4,
      label_size = 4,
      edge_alpha = NULL,
      point_size = 3,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dim3D"
Arguments
                     giotto object
   gobject
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
   dim2_to_use
                     dimension to use on y-axis
   dim3_to_use
                     dimension to use on z-axis
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
                     select subset of cells based on cell IDs
    select_cells
    show_other_cells
                     display not selected cells
   other_cell_color
                     color of not selected cells
    other_point_size
                     size of not selected cells
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
    color_as_factor
                     convert color column to factor
                     color for cells (see details)
    cell_color
    cell_color_code
                     named vector with colors
    show_cluster_center
```

plot center of selected clusters

direction\_test\_CPG 85

```
show_center_label
```

plot label of selected clusters

center\_point\_size

size of center points

label\_size size of labels

edge\_alpha column to use for alpha of the edges

point\_size size of point (cell)

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

show\_legend show legend

#### **Details**

Description of parameters.

#### Value

plotly

## **Examples**

dimPlot3D(gobject)

direction\_test\_CPG

direction\_test\_CPG

### **Description**

shows direction of change

## Usage

```
direction_test(x, min_fdr = 0.05)
```

# Examples

```
direction_test_CPG()
```

86 doHclust

doHclust

doHclust

### **Description**

cluster cells using hierarchical clustering algorithm

#### Usage

```
doHclust(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes_to_use = NULL,
  dim_reduction_to_use = c("cells", "pca", "umap", "tsne"),
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  distance_method = c("pearson", "spearman", "original", "euclidean", "maximum",
  "manhattan", "canberra", "binary", "minkowski"),
agglomeration_method = c("ward.D2", "ward.D", "single", "complete", "average",
    "mcquitty", "median", "centroid"),
  k = 10,
  h = NULL
  name = "hclust",
  return_gobject = TRUE,
  set_seed = T,
  seed_number = 1234
)
```

```
giotto object
gobject
expression_values
                  expression values to use
genes_to_use
                  subset of genes to use
dim_reduction_to_use
                  dimension reduction to use
dim_reduction_name
                  dimensions reduction name
dimensions_to_use
                  dimensions to use
distance_method
                  distance method
{\tt agglomeration\_method}
                  agglomeration method for hclust
k
                  number of final clusters
h
                  cut hierarchical tree at height = h
                  name for hierarchical clustering
return_gobject boolean: return giotto object (default = TRUE)
set_seed
                  set seed
seed_number
                  number for seed
```

doHMRF 87

#### **Details**

Description on how to use Kmeans clustering method.

#### Value

giotto object with new clusters appended to cell metadata

#### See Also

hclust

#### **Examples**

```
doHclust(gobject)
```

doHMRF

doHMRF

#### **Description**

Run HMRF

## Usage

```
doHMRF(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  spatial_network_name = "spatial_network",
  spatial_genes = NULL,
  spatial_dimensions = c("sdimx", "sdimy", "sdimz"),
  dim_reduction_to_use = NULL,
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  name = "test",
  k = 10,
  betas = c(0, 2, 50),
  tolerance = 1e-10,
  zscore = c("none", "rowcol", "colrow"),
  numinit = 100,
  python_path = NULL,
  output_folder = NULL,
  overwrite_output = TRUE
)
```

88 doKmeans

```
spatial_genes
                  spatial genes to use for HMRF
spatial_dimensions
                  select spatial dimensions to use, default is all possible dimensions
dim_reduction_to_use
                  use another dimension reduction set as input
dim_reduction_name
                  name of dimension reduction set to use
dimensions_to_use
                  number of dimensions to use as input
name
                  name of HMRF run
k
                  number of HMRF domains
betas
                  betas to test for
tolerance
                  tolerance
zscore
                  zscore
numinit
                  number of initializations
python_path
                  python path to use
output_folder
                  output folder to save results
overwrite_output
                  overwrite output folder
```

### **Details**

Description of HMRF parameters ...

### Value

Creates a directory with results that can be viewed with viewHMRFresults

#### **Examples**

```
doHMRF(gobject)
```

doKmeans

doKmeans

### **Description**

cluster cells using kmeans algorithm

```
doKmeans(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes_to_use = NULL,
  dim_reduction_to_use = c("cells", "pca", "umap", "tsne"),
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  distance_method = c("original", "pearson", "spearman", "euclidean", "maximum",
```

doKmeans 89

```
"manhattan", "canberra", "binary", "minkowski"),
centers = 10,
iter_max = 100,
nstart = 1000,
algorithm = "Hartigan-Wong",
name = "kmeans",
return_gobject = TRUE,
set_seed = T,
seed_number = 1234
)
```

## **Arguments**

gobject giotto object

expression\_values

expression values to use

genes\_to\_use subset of genes to use

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimensions reduction name

dimensions\_to\_use

dimensions to use

distance\_method

distance method

centers number of final clusters

iter\_max kmeans maximum iterations

nstart kmeans nstart algorithm kmeans algorithm

name name for kmeans clustering

return\_gobject boolean: return giotto object (default = TRUE)

set\_seed set seed

seed\_number number for seed

### **Details**

Description on how to use Kmeans clustering method.

## Value

giotto object with new clusters appended to cell metadata

### See Also

kmeans

## **Examples**

doKmeans(gobject)

90 doLeidenCluster

doLeidenCluster

doLeidenCluster

#### **Description**

cluster cells using a NN-network and the Leiden community detection algorithm

## Usage

```
doLeidenCluster(
  gobject,
  name = "leiden_clus",
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  python_path = NULL,
  resolution = 1,
  weight_col = "weight",
  partition_type = c("RBConfigurationVertexPartition", "ModularityVertexPartition"),
  init_membership = NULL,
  n_iterations = 1000,
  return_gobject = TRUE,
  set_seed = T,
  seed_number = 1234,
  ...
)
```

```
giotto object
gobject
name
                  name for cluster
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use
network_name
python_path
                  specify specific path to python if required
resolution
                  resolution
weight_col
                  weight column to use for edges
partition_type The type of partition to use for optimisation.
init_membership
                  initial membership of cells for the partition
                  number of interations to run the Leiden algorithm. If the number of iterations
n_iterations
                  is negative, the Leiden algorithm is run until an iteration in which there was no
                  improvement.
return_gobject boolean: return giotto object (default = TRUE)
set_seed
                  set seed
seed_number
                  number for seed
```

doLeidenSubCluster 91

#### **Details**

This function is a wrapper for the Leiden algorithm implemented in python, which can detect communities in graphs of millions of nodes (cells), as long as they can fit in memory. See the https://github.com/vtraag/leidenalgleidenalg github page or the https://leidenalg.readthedocs.io/en/stable/index.htmlreadthedocs page for more information.

Partition types available and information:

- RBConfigurationVertexPartition: Implements Reichardt and Bornholdt's Potts model with a configuration null model. This quality function is well-defined only for positive edge weights. This quality function uses a linear resolution parameter.
- Modularity Vertex Partition: Implements modularity. This quality function is well-defined only for positive edge weights. It does *not* use the resolution parameter

Set  $weight\_col = NULL$  to give equal weight (=1) to each edge.

#### Value

giotto object with new clusters appended to cell metadata

#### **Examples**

```
doLeidenCluster(gobject)
```

doLeidenSubCluster

doLeidenSubCluster

### **Description**

Further subcluster cells using a NN-network and the Leiden algorithm

```
doLeidenSubCluster(
  gobject,
  name = "sub_pleiden_clus",
  cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
   = "normalized"),
 hvg_min_perc_cells = 5,
 hvg_mean_expr_det = 1,
 use_all_genes_as_hvg = FALSE,
 min_nr_of_hvg = 5,
 pca_param = list(expression_values = "normalized", scale_unit = T),
 nn_param = list(dimensions_to_use = 1:20),
 k_neighbors = 10,
  resolution = 0.5,
 n_{iterations} = 500,
 python_path = NULL,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
```

92 doLeidenSubCluster

```
return_gobject = TRUE,
verbose = T
)
```

#### **Arguments**

gobject giotto object

name name for new clustering result cluster\_column to subcluster

selected\_clusters

only do subclustering on these clusters

hvg\_param parameters for calculateHVG

hvg\_min\_perc\_cells

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

forces all genes to be HVG and to be used as input for PCA

min\_nr\_of\_hvg minimum number of HVG, or all genes will be used as input for PCA

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

 $k\_neighbors \qquad number \ of \ k \ for \ createNearestNetwork$ 

resolution resolution of Leiden clustering

n\_iterations number of interations to run the Leiden algorithm.

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

return\_gobject boolean: return giotto object (default = TRUE)

verbose verbose

### **Details**

This function performs subclustering using the Leiden algorithm on selected clusters. The systematic steps are:

- 1. subset Giotto object
- 2. identify highly variable genes
- 3. run PCA
- 4. create nearest neighbouring network
- 5. do Leiden clustering

### Value

giotto object with new subclusters appended to cell metadata

doLouvainCluster 93

#### See Also

doLeidenCluster

#### **Examples**

```
doLeidenSubCluster(gobject)
```

doLouvainCluster

doLouvainCluster

### **Description**

cluster cells using a NN-network and the Louvain algorithm.

#### Usage

```
doLouvainCluster(
  gobject,
  version = c("community", "multinet"),
  name = "louvain_clus",
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  python_path = NULL,
  resolution = 1,
  weight_col = NULL,
  gamma = 1,
  omega = 1,
  louv_random = F,
  return_gobject = TRUE,
  set_seed = F,
  seed_number = 1234,
)
```

```
giotto object
gobject
                  implemented version of Louvain clustering to use
version
                  name for cluster
name
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use
network_name
python_path
                  [community] specify specific path to python if required
resolution
                  [community] resolution
                  [multinet] Resolution parameter for modularity in the generalized louvain method.
gamma
                  [multinet] Inter-layer weight parameter in the generalized louvain method.
omega
return_gobject boolean: return giotto object (default = TRUE)
                  set seed
set_seed
seed_number
                  number for seed
```

#### **Details**

Louvain clustering using the community or multinet implementation of the louvain clustering algorithm.

#### Value

giotto object with new clusters appended to cell metadata

#### See Also

doLouvainCluster\_community and doLouvainCluster\_multinet

### **Examples**

```
doLouvainCluster(gobject)
```

```
\label{lower} do Louvain Cluster\_community \\ do Louvain Cluster\_community
```

#### **Description**

cluster cells using a NN-network and the Louvain algorithm from the community module in Python

#### Usage

```
doLouvainCluster_community(
   gobject,
   name = "louvain_clus",
   nn_network_to_use = "sNN",
   network_name = "sNN.pca",
   python_path = NULL,
   resolution = 1,
   weight_col = NULL,
   louv_random = F,
   return_gobject = TRUE,
   set_seed = F,
   seed_number = 1234,
   ...
)
```

```
weight_col weight column to use for edges
```

louv\_random Will randomize the node evaluation order and the community evaluation order

to get different partitions at each call

return\_gobject boolean: return giotto object (default = TRUE)

set\_seed set seed

seed\_number number for seed

#### **Details**

This function is a wrapper for the Louvain algorithm implemented in Python, which can detect communities in graphs of nodes (cells). See the https://python-louvain.readthedocs.io/en/latest/index.htmlreadthedocs page for more information.

Set  $weight\_col = NULL$  to give equal weight (=1) to each edge.

#### Value

giotto object with new clusters appended to cell metadata

#### **Examples**

```
doLouvainCluster_community(gobject)
```

```
doLouvainCluster_multinet
```

doLouvainCluster\_multinet

### **Description**

cluster cells using a NN-network and the Louvain algorithm from the multinet package in R.

```
doLouvainCluster_multinet(
  gobject,
  name = "louvain_clus",
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  gamma = 1,
  omega = 1,
  return_gobject = TRUE,
  set_seed = F,
  seed_number = 1234,
  ...
)
```

96 doLouvainSubCluster

#### **Arguments**

return\_gobject boolean: return giotto object (default = TRUE)

set\_seed set seed

seed\_number number for seed

#### **Details**

See glouvain\_ml from the multinet package in R for more information.

#### Value

giotto object with new clusters appended to cell metadata

### **Examples**

```
doLouvainCluster_multinet(gobject)
```

doLouvainSubCluster doLouvainSubCluster

## Description

subcluster cells using a NN-network and the Louvain algorithm

```
doLouvainSubCluster(
 gobject,
 name = "sub_louvain_clus",
  version = c("community", "multinet"),
  cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
   = "normalized"),
 hvg_min_perc_cells = 5,
 hvg_mean_expr_det = 1,
 use_all_genes_as_hvg = FALSE,
 min_nr_of_hvg = 5,
 pca_param = list(expression_values = "normalized", scale_unit = T),
 nn_param = list(dimensions_to_use = 1:20),
 k_neighbors = 10,
  resolution = 0.5,
```

doLouvainSubCluster 97

```
gamma = 1,
omega = 1,
python_path = NULL,
nn_network_to_use = "sNN",
network_name = "sNN.pca",
return_gobject = TRUE,
verbose = T
)
```

#### **Arguments**

gobject giotto object

name name for new clustering result version version of Louvain algorithm to use

cluster\_column cluster column to subcluster

selected\_clusters

only do subclustering on these clusters

hvg\_param parameters for calculateHVG

 $hvg\_min\_perc\_cells$ 

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

forces all genes to be HVG and to be used as input for PCA

min\_nr\_of\_hvg minimum number of HVG, or all genes will be used as input for PCA

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

 $\label{eq:kneighbors} $$ number of $k$ for createNearestNetwork $$ resolution $$ resolution for community algorithm$ 

gamma gamma omega omega

python\_path specify specific path to python if required

 $nn\_network\_to\_use$ 

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

return\_gobject boolean: return giotto object (default = TRUE)

verbose verbose

#### **Details**

This function performs subclustering using the Louvain algorithm on selected clusters. The systematic steps are:

- 1. subset Giotto object
- 2. identify highly variable genes
- 3. run PCA
- 4. create nearest neighbouring network
- 5. do Louvain clustering

#### Value

giotto object with new subclusters appended to cell metadata

#### See Also

```
doLouvainCluster_multinet and doLouvainCluster_community
```

### **Examples**

```
doLouvainSubCluster(gobject)
```

```
\label{loss} do Louvain SubCluster\_community \\ do Louvain SubCluster\_community
```

### **Description**

subcluster cells using a NN-network and the Louvain community detection algorithm

### Usage

```
doLouvainSubCluster_community(
  gobject,
  name = "sub_louvain_comm_clus",
  cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
    = "normalized"),
  hvg_min_perc_cells = 5,
  hvg_mean_expr_det = 1,
  use_all_genes_as_hvg = FALSE,
  min_nr_of_hvg = 5,
  pca_param = list(expression_values = "normalized", scale_unit = T),
  nn_param = list(dimensions_to_use = 1:20),
  k_neighbors = 10,
  resolution = 0.5,
  python_path = NULL,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  return_gobject = TRUE,
  verbose = T
)
```

```
gobject giotto object

name name for new clustering result

cluster_column cluster column to subcluster

selected_clusters

only do subclustering on these clusters
```

hvg\_param parameters for calculateHVG

hvg\_min\_perc\_cells

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

forces all genes to be HVG and to be used as input for PCA

min\_nr\_of\_hvg minimum number of HVG, or all genes will be used as input for PCA

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

resolution resolution

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

return\_gobject boolean: return giotto object (default = TRUE)

verbose verbose

#### **Details**

This function performs subclustering using the Louvain community algorithm on selected clusters. The systematic steps are:

- 1. subset Giotto object
- 2. identify highly variable genes
- 3. run PCA
- 4. create nearest neighbouring network
- 5. do Louvain community clustering

#### Value

giotto object with new subclusters appended to cell metadata

#### See Also

doLouvainCluster\_community

### **Examples**

doLouvainSubCluster\_community(gobject)

```
\label{lower} do Louvain SubCluster\_multinet \\ do Louvain SubCluster\_multinet
```

## Description

subcluster cells using a NN-network and the Louvain multinet detection algorithm

### Usage

```
doLouvainSubCluster_multinet(
  gobject,
 name = "sub_louvain_mult_clus",
 cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
   = "normalized"),
 hvg_min_perc_cells = 5,
 hvg_mean_expr_det = 1,
 use_all_genes_as_hvg = FALSE,
 min_nr_of_hvg = 5,
 pca_param = list(expression_values = "normalized", scale_unit = T),
 nn_param = list(dimensions_to_use = 1:20),
 k_neighbors = 10,
  gamma = 1,
  omega = 1,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 return_gobject = TRUE,
  verbose = T
)
```

```
giotto object
gobject
name
                  name for new clustering result
cluster_column cluster column to subcluster
selected_clusters
                  only do subclustering on these clusters
                  parameters for calculateHVG
hvg_param
hvg_min_perc_cells
                  threshold for detection in min percentage of cells
hvg_mean_expr_det
                  threshold for mean expression level in cells with detection
use_all_genes_as_hvg
                  forces all genes to be HVG and to be used as input for PCA
                  minimum number of HVG, or all genes will be used as input for PCA
min_nr_of_hvg
                  parameters for runPCA
pca_param
```

doRandomWalkCluster 101

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

gamma gamma omega omega

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

return\_gobject boolean: return giotto object (default = TRUE)

verbose verbose

python\_path specify specific path to python if required

### **Details**

This function performs subclustering using the Louvain multinet algorithm on selected clusters. The systematic steps are:

- 1. subset Giotto object
- 2. identify highly variable genes
- 3. run PCA
- 4. create nearest neighbouring network
- 5. do Louvain multinet clustering

### Value

giotto object with new subclusters appended to cell metadata

#### See Also

doLouvainCluster\_multinet

## **Examples**

doLouvainSubCluster\_multinet(gobject)

 $do Random Walk Cluster \qquad do Random Walk Cluster$ 

## Description

Cluster cells using a random walk approach.

102 doRandomWalkCluster

#### Usage

```
doRandomWalkCluster(
  gobject,
  name = "random_walk_clus",
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  walk_steps = 4,
  walk_clusters = 10,
  walk_weights = NA,
  return_gobject = TRUE,
  set_seed = F,
  seed_number = 1234,
  ...
)
```

### **Arguments**

```
gobject
                 giotto object
                 name for cluster
name
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
network_name
                 name of NN network to use
walk_steps
                 number of walking steps
                 number of final clusters
walk_clusters
                 cluster column defining the walk weights
walk_weights
return_gobject boolean: return giotto object (default = TRUE)
set\_seed
                 set seed
seed_number
                 number for seed
```

### **Details**

See cluster\_walktrap function from the igraph package in R for more information.

## Value

giotto object with new clusters appended to cell metadata

## **Examples**

```
doRandomWalkCluster(gobject)
```

doSNNCluster 103

doSNNCluster doSNNCluster

## Description

Cluster cells using a SNN cluster approach.

### Usage

```
doSNNCluster(
  gobject,
  name = "sNN_clus",
  nn_network_to_use = "kNN",
  network_name = "kNN.pca",
  k = 20,
  eps = 4,
  minPts = 16,
  borderPoints = TRUE,
  return_gobject = TRUE,
  set_seed = F,
  seed_number = 1234,
  ...
)
```

### Arguments

gobject giotto object name name for cluster

nn\_network\_to\_use

type of NN network to use (only works on kNN)

network\_name name of kNN network to use

k Neighborhood size for nearest neighbor sparsification to create the shared NN

graph.

eps Two objects are only reachable from each other if they share at least eps nearest

neighbors.

minPts minimum number of points that share at least eps nearest neighbors for a point

to be considered a core points.

borderPoints should borderPoints be assigned to clusters like in DBSCAN?

return\_gobject boolean: return giotto object (default = TRUE)

set\_seed set seed

seed\_number number for seed

### **Details**

See sNNclust from dbscan package

#### Value

giotto object with new clusters appended to cell metadata

### **Examples**

```
doSNNCluster(gobject)
```

## Description

smooth gene expression over a defined spatial grid

# Usage

```
do_spatial_grid_averaging(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_grid_name = "spatial_grid",
  min_cells_per_grid = 4
)
```

# Arguments

```
gobject giotto object
expression_values
gene expression values to use
subset_genes subset of genes to use
spatial_grid_name
name of spatial grid to use
min_cells_per_grid
minimum number of cells to consider a grid
```

## Value

matrix with smoothened gene expression values based on spatial grid

## **Examples**

```
do_spatial_grid_averaging(gobject)
```

```
\begin{tabular}{ll} $do\_spatial\_knn\_smoothing \\ & do\_spatial\_knn\_smoothing \\ \end{tabular}
```

## Description

smooth gene expression over a kNN spatial network

### Usage

```
do_spatial_knn_smoothing(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_network_name = "spatial_network",
  b = NULL
)
```

### **Arguments**

### Details

This function will smoothen the gene expression values per cell according to its neighbors in the selected spatial network.

b is a smoothening factor that defaults to 1 - 1/k, where k is the median number of k-neighbors in the selected spatial network. Setting b = 0 means no smoothing and b = 1 means no contribution from its own expression.

#### Value

matrix with smoothened gene expression values based on kNN spatial network

## **Examples**

```
do_spatial_knn_smoothing(gobject)
```

### **Description**

converts data.table to matrix

### Usage

```
dt_to_matrix(x)
```

### **Examples**

```
dt_to_matrix(x)
```

 ${\tt enrichSpatialCorGroups}$ 

enrichSpatialCorGroups

# Description

Create enrichment scores based on the metagene expression of the spatially correlated gene groups.

#### Usage

```
enrichSpatialCorGroups(
  gobject,
  spatCorObject,
  expression_values = c("normalized", "scaled", "custom"),
  use_clus_name = NULL,
  select_clusters = NULL,
  name = "spatclus_enr",
  convert_enrich_to_cluster = FALSE,
  enrich_to_cluster_name = "enrich_cluster",
  return_gobject = TRUE
)
```

exportGiottoViewer 107

#### **Details**

This function calculates the metagene expression for each identified cluster of spatially correlated genes (use\_clus\_name) and provides a new enrichment score (name) that can be visualized. Additionally, these spatial gene correlation enrichment scores can be converted into clusters by selecting the highest z-score per cell after z-scoring first columns and then rows.

#### Value

giotto object

### **Examples**

```
enrichSpatialCorGroups(gobject)
```

exportGiottoViewer

*exportGiottoViewer* 

## **Description**

compute highly variable genes

```
exportGiottoViewer(
  gobject,
  output_directory = NULL,
  spat_enr_names = NULL,
  factor_annotations,
  numeric_annotations,
  dim_reductions,
  dim_reduction_names,
  expression_values = c("scaled", "normalized", "custom"),
  dim_red_rounding = NULL,
  dim_red_rescale = c(-20, 20),
  expression_rounding = 2,
  overwrite_dir = T,
  verbose = T
```

#### **Arguments**

```
gobject
                  giotto object
output_directory
                  directory where to save the files
spat_enr_names spatial enrichment results to include for annotations
factor_annotations
                  giotto cell annotations to view as factor
numeric_annotations
                  giotto cell annotations to view as numeric
dim_reductions high level dimension reductions to view
dim_reduction_names
                  specific dimension reduction names
expression_values
                  expression values to use in Viewer
dim_red_rounding
                  numerical indicating how to round the coordinates
dim_red_rescale
                  numericals to rescale the coordinates
expression_rounding
                  numerical indicating how to round the expression data
                  overwrite files in the directory if it already existed
overwrite_dir
verbose
                  be verbose
```

#### **Details**

Giotto Viewer expects the results from Giotto Analyzer in a specific format, which is provided by this function. To include enrichment results from createSpatialEnrich include the provided spatial enrichment name (default PAGE or rank) and add the gene signature names (.e.g cell types) to the numeric annotations parameter.

## Value

writes the necessary output to use in Giotto Viewer

## **Examples**

```
exportGiottoViewer(gobject)
```

```
expr Only Cell Cell communication Scores \\ expr Only Cell Cell communication Scores
```

### **Description**

Cell-Cell communication scores based on expression only

extended\_gini\_fun 109

#### Usage

```
exprOnlyCellCellcommunicationScores(
  gobject,
  cluster_column = "cell_types",
  random_iter = 100,
  gene_set_1,
  gene_set_2,
  log2FC_addendum = 0.1,
  verbose = T
)
```

#### **Arguments**

```
gobject giotto object to use

cluster_column cluster column with cell type information

random_iter number of iterations

gene_set_1 first specific gene set from gene pairs

gene_set_2 second specific gene set from gene pairs

log2FC_addendum

addendum to add when calculating log2FC

verbose verbose
```

#### **Details**

Statistical framework to identify if pairs of genes (such as ligand-receptor combinations) are expressed at higher levels than expected based on a reshuffled null distribution of gene expression values, without considering the spatial position of cells. More details will follow soon.

### Value

Cell-Cell communication scores for gene pairs based on expression only

### **Examples**

```
exprOnlyCellCellcommunicationScores(gobject)
```

```
extended_gini_fun extended_gini_fun
```

# Description

calculate gini coefficient on a minimum length vector

### Usage

```
extended_gini_fun(x, weights = rep(1, length = length(x)), minimum_length = 16)
```

### Value

gini coefficient

110 fDataDT

```
{\tt extractNearestNetwork} \ \ {\it extractNearestNetwork}
```

## **Description**

Extracts a NN-network from a Giotto object

## Usage

```
extractNearestNetwork(
  gobject,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  output = c("igraph", "data.table")
)
```

## Arguments

### Value

igraph or data.table object

### **Examples**

extractNearestNetwork(gobject)

 ${\sf fDataDT}$ 

fDataDT

### **Description**

show gene metadata

# Usage

```
fDataDT(gobject)
```

## **Arguments**

gobject giotto object

### Value

data.table with gene metadata

filterCombinations 111

#### **Examples**

```
pDataDT(gobject)
```

filterCombinations

*filterCombinations* 

#### **Description**

Shows how many genes and cells are lost with combinations of thresholds.

#### Usage

```
filterCombinations(
  gobject,
  expression_values = c("raw", "normalized", "scaled", "custom"),
  expression_thresholds = c(1, 2),
  gene_det_in_min_cells = c(5, 50),
  min_det_genes_per_cell = c(200, 400),
  scale_x_axis = "identity",
  x_axis_offset = 0,
  scale_y_axis = "identity",
  y_axis_offset = 0,
  show_plot = TRUE
)
```

## **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
expression_thresholds
                  all thresholds to consider a gene expressed
gene_det_in_min_cells
                  minimum number of cells that should express a gene to consider that gene fur-
                  ther
min_det_genes_per_cell
                  minimum number of expressed genes per cell to consider that cell further
scale_x_axis
                  ggplot transformation for x-axis (e.g. log2)
x_axis_offset
                 x-axis offset to be used together with the scaling transformation
                  ggplot transformation for y-axis (e.g. log2)
scale_y_axis
y_axis_offset
                  y-axis offset to be used together with the scaling transformation
                  show plot
show_plot
```

### **Details**

Creates a scatterplot that visualizes the number of genes and cells that are lost with a specific combination of a gene and cell threshold given an arbitrary cutoff to call a gene expressed. This function can be used to make an informed decision at the filtering step with filterGiotto.

112 filterCPGscores

#### Value

list of data.table and ggplot object

### **Examples**

```
filterCombinations(gobject)
```

filterCPGscores

filterCPGscores

#### **Description**

visualize Cell Proximity Gene enrichment scores

### Usage

```
filterCPGscores(
   CPGscore,
   min_cells = 5,
   min_fdr = 0.05,
   min_spat_diff = 0.2,
   min_log2_fc = 0.5,
   keep_int_duplicates = TRUE,
   direction = c("both", "up", "down")
)
```

### **Arguments**

```
min_cells min number of cells threshold
min_fdr false_discovery threshold
min_spat_diff spatial difference threshold
min_log2_fc min log2 fold-change
keep_int_duplicates
keep both cell_A-cell_B and cell_B-cell_A
direction expression changes to keep
method visualization method
```

## **Details**

This function filters the output from getCellProximityGeneScores based on false-discovery rate, minimum absolute difference, minimum log fold-change and direction of change.

#### Value

Gene to gene scores in data.table format

```
filterCPGscores(CPGscore)
```

filterDistributions 113

```
filterDistributions filterDistributions
```

### **Description**

show gene or cell distribution after filtering on expression threshold

## Usage

```
filterDistributions(
  gobject,
  expression_values = c("raw", "normalized", "scaled", "custom"),
  expression_threshold = 1,
  detection = c("genes", "cells"),
  plot_type = c("histogram", "violin"),
  nr_bins = 30,
  fill_color = "lightblue",
  scale_axis = "identity",
  axis_offset = 0,
  show_plot = TRUE
)
```

## Arguments

```
gobject
                  giotto object
expression_values
                  expression values to use
expression_threshold
                  threshold to consider a gene expressed
detection
                  consider genes or cells
plot_type
                  type of plot
nr_bins
                  number of bins for histogram plot
fill_color
                  fill color for plots
scale_axis
                  ggplot transformation for axis (e.g. log2)
                  offset to be used together with the scaling transformation
axis_offset
show_plot
                  show plot
```

## Value

ggplot object

```
filterDistributions(gobject)
```

114 filterGiotto

filterGiotto

filterGiotto

## Description

filter Giotto object based on expression threshold

### Usage

```
filterGiotto(
  gobject,
  expression_values = c("raw", "normalized", "scaled", "custom"),
  expression_threshold = 1,
  gene_det_in_min_cells = 100,
  min_det_genes_per_cell = 100,
  verbose = F
)
```

# Arguments

```
gobject giotto object

expression_values

expression values to use

expression_threshold

threshold to consider a gene expressed

gene_det_in_min_cells

minimum # of cells that need to express a gene

min_det_genes_per_cell

minimum # of genes that need to be detected in a cell

verbose

verbose
```

### **Details**

The function filterCombinations can be used to explore the effect of different parameter values.

## Value

giotto object

```
filterGiotto(gobject)
```

findGiniMarkers 115

findGiniMarkers

findGiniMarkers

#### **Description**

Identify marker genes for selected clusters based on gini detection and expression scores.

#### Usage

```
findGiniMarkers(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  group_1 = NULL,
  group_2 = NULL,
  min_expr_gini_score = 0.5,
  min_det_gini_score = 0.5,
  detection_threshold = 0,
  rank_score = 1
)
```

## Arguments

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
subset_clusters
                  selection of clusters to compare
group_1
                  group 1 cluster IDs from cluster_column for pairwise comparison
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
min_expr_gini_score
                  filter on minimum gini coefficient for expression
min_det_gini_score
                  filter on minimum gini coefficient for detection
detection_threshold
                  detection threshold for gene expression
                  rank scores to include
rank_score
```

#### **Details**

Detection of marker genes using the <a href="https://en.wikipedia.org/wiki/Gini\_coefficientginic

- 1. calculate average expression per cluster
- 2. calculate detection fraction per cluster
- 3. calculate gini-coefficient for av. expression values over all clusters

- 4. calculate gini-coefficient for detection fractions over all clusters
- 5. convert gini-scores to rank scores
- 6. for each gene create combined gini score = av. expr gini x detection gini
- 7. for each gene create and sort on combined rank score = expr rank x detection rank

As a results "top gini" genes are genes that are very selectivily expressed in a specific cluster, however not always expressed in all cells of that cluster. In other words highly specific, but not necessarily sensitive at the single-cell level.

To perform differential expression between cluster groups you need to specificy cluster IDs to the parameters *group\_1* and *group\_2*.

#### Value

data.table with marker genes

#### **Examples**

```
findGiniMarkers(gobject)
```

#### **Description**

Identify marker genes for all clusters in a one vs all manner based on gini detection and expression scores.

#### Usage

```
findGiniMarkers_one_vs_all(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  min_expr_gini_score = 0.5,
  min_det_gini_score = 0.5,
  detection_threshold = 0,
  min_genes = 10,
  verbose = TRUE
)
```

```
gobject giotto object
expression_values
gene expression values to use
cluster_column clusters to use
subset_clusters
selection of clusters to compare
```

findMarkers 117

```
min_expr_gini_score
filter on minimum gini coefficient on expression

min_det_gini_score
filter on minimum gini coefficient on detection

detection_threshold
detection threshold for gene expression

min_genes minimum genes to keep per cluster, overrides pval and logFC

verbose be verbose
```

### Value

data.table with marker genes

#### See Also

findGiniMarkers

### **Examples**

```
findGiniMarkers_one_vs_all(gobject)
```

findMarkers findMarkers

## Description

Identify marker genes for selected clusters.

# Usage

```
findMarkers(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  method = c("scran", "gini", "mast"),
  subset_clusters = NULL,
  group_1 = NULL,
  group_2 = NULL,
  min_expr_gini_score = 0.5,
  min_det_gini_score = 0.5,
  detection_threshold = 0,
  rank\_score = 1,
  group_1_name = NULL,
  group_2_name = NULL,
  adjust_columns = NULL,
)
```

118 findMarkers

#### **Arguments**

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
method
                  method to use to detect differentially expressed genes
subset_clusters
                  selection of clusters to compare
                  group 1 cluster IDs from cluster_column for pairwise comparison
group_1
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
min_expr_gini_score
                  gini: filter on minimum gini coefficient for expression
min_det_gini_score
                  gini: filter minimum gini coefficient for detection
detection_threshold
                  gini: detection threshold for gene expression
rank_score
                  gini: rank scores to include
                  mast: custom name for group_1 clusters
group_1_name
                  mast: custom name for group_2 clusters
group_2_name
adjust_columns mast: column in pDataDT to adjust for (e.g. detection rate)
                  additional parameters for the findMarkers function in scran or zlm function in
                  MAST
```

#### **Details**

Wrapper for all individual functions to detect marker genes for clusters.

#### Value

data.table with marker genes

## See Also

 $find Scran Markers, find Gini Markers \ and \ find Mast Markers$ 

#### **Examples**

findMarkers(gobject)

```
find {\it Markers\_one\_vs\_all} \\ find {\it Markers\_one\_vs\_all}
```

### **Description**

Identify marker genes for all clusters in a one vs all manner.

#### Usage

```
findMarkers_one_vs_all(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 cluster_column,
  subset_clusters = NULL,
 method = c("scran", "gini", "mast"),
 pval = 0.01,
 logFC = 0.5,
 min\_genes = 10,
 min_expr_gini_score = 0.5,
 min_det_gini_score = 0.5,
 detection_threshold = 0,
  rank_score = 1,
 adjust_columns = NULL,
 verbose = TRUE,
)
```

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
subset_clusters
                  selection of clusters to compare
                  method to use to detect differentially expressed genes
method
pval
                  scran & mast: filter on minimal p-value
logFC
                  scan & mast: filter on logFC
min_genes
                  minimum genes to keep per cluster, overrides pval and logFC
min_expr_gini_score
                  gini: filter on minimum gini coefficient for expression
min_det_gini_score
                  gini: filter minimum gini coefficient for detection
detection_threshold
                  gini: detection threshold for gene expression
                  gini: rank scores to include
rank_score
adjust_columns mast: column in pDataDT to adjust for (e.g. detection rate)
```

120 findMastMarkers

```
verbose be verbose
... additional parameters for the findMarkers function in scran or zlm function in MAST
```

#### **Details**

Wrapper for all one vs all functions to detect marker genes for clusters.

#### Value

data.table with marker genes

#### See Also

```
find Scran Markers\_one\_vs\_all, find Gini Markers\_one\_vs\_all \ and \ find Mast Markers\_one\_vs\_all \ and \ find Markers\_one\_v
```

### **Examples**

```
{\tt findMarkers\_one\_vs\_all(gobject)}
```

findMastMarkers

findMastMarkers

### **Description**

Identify marker genes for selected clusters based on the MAST package.

#### Usage

```
findMastMarkers(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  group_1 = NULL,
  group_1_name = NULL,
  group_2 = NULL,
  group_2_name = NULL,
  adjust_columns = NULL,
  ...
)
```

```
gobject giotto object
expression_values
gene expression values to use

cluster_column clusters to use
group_1 group 1 cluster IDs from cluster_column for pairwise comparison
group_1_name custom name for group_1 clusters
group_2 group 2 cluster IDs from cluster_column for pairwise comparison
```

```
group_2_name custom name for group_2 clusters
adjust_columns column in pDataDT to adjust for (e.g. detection rate)
... additional parameters for the zlm function in MAST
```

#### **Details**

This is a minimal convenience wrapper around the zlm from the MAST package to detect differentially expressed genes.

#### Value

data.table with marker genes

#### **Examples**

```
findMastMarkers(gobject)
```

```
findMastMarkers_one_vs_all findMastMarkers_one_vs_all
```

## Description

Identify marker genes for all clusters in a one vs all manner based on the MAST package.

### Usage

```
findMastMarkers_one_vs_all(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  adjust_columns = NULL,
  pval = 0.001,
  logFC = 1,
  min_genes = 10,
  verbose = TRUE,
  ...
)
```

```
gobject giotto object
expression_values
gene expression values to use
cluster_column clusters to use
subset_clusters
selection of clusters to compare
adjust_columns column in pDataDT to adjust for (e.g. detection rate)
pval filter on minimal p-value
```

122 findScranMarkers

```
logFC filter on logFC
min_genes minimum genes to keep per cluster, overrides pval and logFC
verbose be verbose
... additional parameters for the zlm function in MAST
```

#### Value

data.table with marker genes

#### See Also

findMastMarkers

### **Examples**

```
findMastMarkers_one_vs_all(gobject)
```

findScranMarkers findScranMarkers

#### **Description**

Identify marker genes for all or selected clusters based on scran's implementation of findMarkers.

### Usage

```
findScranMarkers(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  group_1 = NULL,
  group_2 = NULL,
  ...
)
```

```
gobject giotto object
expression_values
gene expression values to use

cluster_column clusters to use
subset_clusters
selection of clusters to compare

group_1 group 1 cluster IDs from cluster_column for pairwise comparison
group_2 group 2 cluster IDs from cluster_column for pairwise comparison
additional parameters for the findMarkers function in scran
```

#### **Details**

This is a minimal convenience wrapper around the findMarkers function from the scran package.

To perform differential expression between cluster groups you need to specificy cluster IDs to the parameters *group\_1* and *group\_2*.

### Value

data.table with marker genes

### **Examples**

```
findScranMarkers(gobject)
```

```
find Scran Markers\_one\_vs\_all \\ find Scran Markers\_one\_vs\_all
```

### **Description**

Identify marker genes for all clusters in a one vs all manner based on scran's implementation of findMarkers.

#### Usage

```
findScranMarkers_one_vs_all(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  pval = 0.01,
  logFC = 0.5,
  min_genes = 10,
  verbose = TRUE,
  ...
)
```

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
subset_clusters
                  subset of clusters to use
pval
                  filter on minimal p-value
logFC
                  filter on logFC
                  minimum genes to keep per cluster, overrides pval and logFC
min_genes
verbose
                  be verbose
                  additional parameters for the findMarkers function in scran
. . .
```

find\_grid\_x

#### Value

data.table with marker genes

## See Also

findScranMarkers

## **Examples**

findScranMarkers\_one\_vs\_all(gobject)

find\_grid\_2D

find\_grid\_2D

# Description

find grid location in 2D

## Usage

```
find_grid_2D(grid_DT, x_loc, y_loc)
```

find\_grid\_3D

find\_grid\_3D

# Description

find grid location in 3D

# Usage

```
find_grid_3D(grid_DT, x_loc, y_loc, z_loc)
```

find\_grid\_x

find\_grid\_x

# Description

find grid location on x-axis

## Usage

```
find_grid_x(grid_DT, x_loc)
```

find\_grid\_y

find\_grid\_y

 $find\_grid\_y$ 

## Description

find grid location on y-axis

## Usage

```
find_grid_y(grid_DT, y_loc)
```

 ${\tt find\_grid\_z}$ 

 $find\_grid\_z$ 

## Description

find grid location on z-axis

# Usage

```
find_grid_z(grid_DT, z_loc)
```

fish\_function

fish\_function

## Description

perform fisher exact test

## Usage

```
fish_function(x_to, x_from)
```

fish\_function2

fish\_function2

# Description

perform fisher exact test

## Usage

```
fish_function2(A, B, C, D)
```

FSV\_show

FSV\_show

FSV\_show

# Description

Visualize spatial varible genes caculated by spatial\_DE

# Usage

```
FSV_show(
  results,
  ms_results = NULL,
  size = c(4, 2, 1),
  color = c("blue", "green", "red"),
  sig_alpha = 0.5,
  unsig_alpha = 0.5
)
```

## **Arguments**

results results caculated by spatial\_DE

ms\_results ms\_results caculated by spatial\_DE

size indicate different levels of qval

color indicate different SV features

sig\_alpha transparency of significant genes

unsig\_alpha transparency of unsignificant genes

# **Details**

Description of parameters.

## Value

nothing

```
FSV_show(results)
```

GenePattern\_show 127

GenePattern\_show

GenePattern\_show

## Description

Visualize genes distribution patterns calculated by spatial\_AEH

### Usage

```
GenePattern_show(
  gobject = NULL,
  AEH_results = NULL,
  sdimx = NULL,
  sdimy = NULL,
  point_size = 3,
  point_alpha = 1,
  low_color = "blue",
  mid_color = "white",
  high_color = "red",
  midpoint = 0
)
```

### **Arguments**

gobject giotto object results from spatial\_AEH AEH\_results sdimx x axis of spatial locus sdimy y axis of spatial locus point\_size size of points to indicate cells transparency of points to indicate cells point\_alpha low\_color color to indicate low score level color to indicate middle score level mid\_color high\_color color to indicate high score level point to set mid\_color midpoint

## Details

Description of parameters.

### Value

nothing

```
GenePattern_show(gobject,AEH_results)
```

128 general\_save\_function

```
general_save_function general_save_function
```

### **Description**

Function to automatically save plots to directory of interest

### Usage

```
general_save_function(
   gobject,
   plot_object,
   save_dir = NULL,
   save_folder = NULL,
   save_name = NULL,
   default_save_name = "giotto_plot",
   save_format = c("png", "tiff", "pdf", "svg"),
   show_saved_plot = F,
   base_width = NULL,
   base_height = NULL,
   base_aspect_ratio = NULL,
   units = NULL,
   dpi = NULL,
   ...
)
```

# Arguments

```
gobject
                  giotto object
                  non-ggplot object to plot
plot_object
save_dir
                  directory to save to
save_folder
                  folder in save_dir to save to
                  name of plot
save_name
save_format
                  format (e.g. png, tiff, pdf, ...)
show_saved_plot
                  load & display the saved plot
base_width
                  width
base_height
                  height
base_aspect_ratio
                  aspect ratio
units
                  units
dpi
                  Plot resolution
```

```
general_save_function(gobject)
```

get10Xmatrix 129

get10Xmatrix

get10Xmatrix

### **Description**

This function creates an expression matrix from a 10X structured folder

### Usage

```
get10Xmatrix(path_to_data)
```

### **Arguments**

```
path_to_data path to the 10X folder
```

#### **Details**

A typical 10X folder is named raw\_feature\_bc\_matrix or raw\_feature\_bc\_matrix. It has 3 files:

- barcodes
- features.tsv.gz
- matrix.mtx.gz

#### Value

expression matrix from 10X

### **Examples**

```
get10Xmatrix(10Xmatrix)
```

```
getCellProximityGeneScores
```

get Cell Proximity Gene Scores

### **Description**

Compute cell-cell interaction enrichment (observed vs expected)

### Usage

```
getCellProximityGeneScores(
  gobject,
  spatial_network_name = "spatial_network",
  cluster_column = "louvain_clus.1",
  selected_genes = NULL,
  expression_values = c("normalized", "scaled", "custom"),
  do_diff_test = TRUE,
  diff_test = c("t.test", "wilcox"),
  minimum_unique_cells = NA,
```

```
fold_change_addendum = 0.1,
  in_two_directions = TRUE,
  exclude_selected_cells_from_test = F,
  verbose = T
)
```

### **Arguments**

```
gobject
                  giotto object
spatial_network_name
                  name of spatial network to use
cluster_column name of column to use for clusters
selected_genes selection of genes to perform calculations for
expression_values
                  expression values to use
do_diff_test
                  perform differential test
diff_test
                  which differential expression test
minimum_unique_cells
                  minimum number of cells needed to proceed
fold_change_addendum
                  constant to add when calculating log2 fold-change
in_two_directions
                  shows enrichment in both directions: cell1-cell2, cell2-cell1
exclude_selected_cells_from_test
                  exclude certain cells from test
verbose
                  verbose
```

#### **Details**

Function to calculate if genes are differentially expressed in cell types when they interact (according to physical proximity) with other cell types. The results data.table contains the following columns:

- genes: All or selected list of tested genes
- cell\_expr\_1: average gene expression in cell type 1 from unified\_int cell-cell interaction
- cell\_expr\_2: average gene expression in cell type 2 from unified\_int cell-cell interaction
- comb\_expr: combined average gene expression in cell type 1 and 2 from unified\_int cell-cell interaction
- all\_cell\_expr\_1: average gene expression for all cells from cell type 1
- all\_cell\_expr\_2: average gene expression for all cells from cell type 2
- all\_comb\_expr: combined average gene expression for all cells from cell type 1 and 2
- pval\_1: p-value from test between interacting cells and all cells from cell type 1
- pval\_2: p-value from test between interacting cells and all cells from cell type 2
- cell\_type\_1: first cell type of cell-cell interaction
- cell\_type\_2: second cell type of cell-cell interaction
- interaction: the cell-cell interaction, based on physical proximity
- nr\_1: number of cell type 1 in the unified cell-cell interaction
- nr\_2: number of cell type 2 in the unified cell-cell interaction

getClusterSimilarity 131

- all\_nr\_1: number of all cell type 1 in the whole dataset
- all\_nr\_2: number of all cell type 2 in the whole dataset
- diff\_spat: difference between comb\_expr and all\_comb\_expr
- diff\_spat\_1: difference between cell\_expr\_1 and all\_cell\_expr\_1
- diff\_spat\_2: difference between cell\_expr\_1 and all\_cell\_expr\_1
- log2fc\_spat\_1: fold-change of diff\_spat\_1
- log2fc\_spat\_2: fold-change of diff\_spat\_2
- log2fc\_spat: fold-change of diff\_spat
- type\_int: type of interaction
- unified\_int: interaction with alphabetically sorted cell type 1 and cell type 2
- unif\_int\_rank: 1 or 2
- fdr\_1: fdr from test between interacting cells and all cells from cell type 1
- fdr\_2: fdr from test between interacting cells and all cells from cell type 2

#### Value

Cell Proximity Gene scores (CPGscores) in data.table format

#### **Examples**

```
getCellProximityGeneScores(gobject)
```

```
getClusterSimilarity
```

### **Description**

Creates data.table with pairwise correlation scores between each cluster.

### Usage

```
getClusterSimilarity(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  cor = c("pearson", "spearman")
)
```

```
gobject giotto object
expression_values
expression values to use
cluster_column name of column to use for clusters
cor correlation score to calculate distance
```

132 getDendrogramSplits

#### **Details**

Creates data.table with pairwise correlation scores between each cluster and the group size (# of cells) for each cluster. This information can be used together with mergeClusters to combine very similar or small clusters into bigger clusters.

### Value

data.table

### **Examples**

```
getClusterSimilarity(gobject)
```

```
getDendrogramSplits getDendrogramSplits
```

### **Description**

Split dendrogram at each node and keep the leave (label) information..

### Usage

```
getDendrogramSplits(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  cor = c("pearson", "spearman"),
  distance = "ward.D",
  h = NULL,
  h_color = "red",
  show_dend = TRUE,
  verbose = TRUE
)
```

```
gobject
                  giotto object
expression_values
                  expression values to use
cluster_column name of column to use for clusters
                  correlation score to calculate distance
cor
distance
                  distance method to use for hierarchical clustering
h
                  height of horizontal lines to plot
h_color
                  color of horizontal lines
                  show dendrogram
show_dend
verbose
                  be verbose
```

getDistinctColors 133

#### **Details**

Creates a data.table with three columns and each row represents a node in the dendrogram. For each node the height of the node is given together with the two subdendrograms. This information can be used to determine in a hierarchical manner differentially expressed marker genes at each node.

#### Value

data.table object

### **Examples**

```
getDendrogramSplits(gobject)
```

getDistinctColors

getDistinctColors

### **Description**

Returns a number of distint colors based on the RGB scale

#### Usage

```
getDistinctColors(n)
```

### **Arguments**

n

number of colors wanted

## Value

number of distinct colors

 ${\tt getGeneToGeneScores}$ 

getGeneToGeneScores

### **Description**

Compute gene-gene enrichment scores.

### Usage

```
getGeneToGeneScores(
   CPGscore,
   selected_genes = NULL,
   specific_genes_1 = NULL,
   specific_genes_2 = NULL,
   min_cells = 5,
   min_fdr = 0.05,
   min_spat_diff = 0.2,
   min_log2_fc = 0.5,
```

```
direction = c("both", "up", "down"),
fold_change_addendum = 0.1,
verbose = TRUE
)
```

## **Arguments**

```
CPGscore
                  CPGscore, output from getCellProximityGeneScores()
selected_genes select subset of genes
specific_genes_1
                  specific source genes (see details)
specific_genes_2
                  specific target genes (see details)
                  min number of cells threshold
min_cells
                  spatial difference threshold
min_spat_diff
min_log2_fc
                  log2 fold-change threshold
direction
                  up or downregulation or both
fold_change_addendum
                  constant to add when calculating log2 fold-change
verbose
                  verbose
min_pval
                  p-value threshold
```

#### **Details**

This converts the single gene cell proximity scores into pairwise combinations of genes, which allows you to determine if 2 genes are differentially expressed in interacting cell types.

## Value

Gene to gene scores in data.table format

### **Examples**

```
getGeneToGeneScores(CPGscore)
```

```
{\it get\_cell\_to\_cell\_sorted\_name\_conversion} \\ {\it get\_cell\_to\_cell\_sorted\_name\_conversion}
```

#### **Description**

creates unified cell-cell interaction names

### Usage

```
get_cell_to_cell_sorted_name_conversion(all_cell_types)
```

```
get_cell_to_cell_sorted_name_conversion()
```

```
{\it get\_interaction\_gene\_enrichment} \\ {\it get\_interaction\_gene\_enrichment}
```

### **Description**

Computes gene enrichment between all interactions

## Usage

```
get_interaction_gene_enrichment(
  spatial_network,
  unified_int_col = "unified_int",
  source_col = "source_clus",
  source_IDs = "from",
  neighb_col = "neighb_clus",
  neighb_IDs = "to",
  expression_matrix,
  cell_annotation,
  annotation_ID = "uniq_ID",
  cell_type_col,
  do_diff_test = T,
  diff_test = c("t.test", "wilcox"),
  minimum_unique_cells = NA,
  exclude_selected_cells_from_test = T,
  verbose = T
)
```

### **Examples**

```
get_interaction_gene_enrichment()
```

```
{\it get\_specific\_interaction\_gene\_enrichment} \\ {\it get\_specific\_interaction\_gene\_enrichment}
```

### **Description**

Computes gene enrichment between specified interaction

### Usage

```
get_specific_interaction_gene_enrichment(
  sub_spatial_network,
  source_col = "source_clus",
  source_IDs = "from",
  neighb_col = "neighb_clus",
  neighb_IDs = "to",
  expression_matrix,
```

136 ggplot\_save\_function

```
interaction_name = "to_specify",
  cell_annotation,
  annotation_ID = "uniq_ID",
  cell_type_col,
  do_diff_test = T,
  diff_test = c("t.test", "wilcox"),
  minimum_unique_cells = NA,
  exclude_selected_cells_from_test = T
)
```

## Examples

```
get_specific_interaction_gene_enrichment()
```

```
ggplot_save_function ggplot_save_function
```

### **Description**

Function to automatically save plots to directory of interest

## Usage

```
ggplot_save_function(
  gobject,
  plot_object,
  save_dir = NULL,
  save_folder = NULL,
  save_name = NULL,
  default_save_name = "giotto_plot",
  save_format = NULL,
  show_saved_plot = F,
  ncol = 1,
  nrow = 1,
  scale = 1,
  base_width = NULL,
  base_height = NULL,
  base_aspect_ratio = NULL,
  units = NULL,
  dpi = NULL,
  limitsize = TRUE,
)
```

```
gobject giotto object
plot_object ggplot object to plot
save_dir directory to save to
save_folder folder in save_dir to save to
```

giotto-class 137

save\_name name of plot

save\_format format (e.g. png, tiff, pdf, ...)

show\_saved\_plot

load & display the saved plot

ncol number of columns nrow number of rows

scale scale
base\_width width
base\_height height
base\_aspect\_ratio

aspect ratio

units units

dpi Plot resolution

limitsize When TRUE (the default), ggsave will not save images larger than 50x50 inches,

to prevent the common error of specifying dimensions in pixels.

#### See Also

```
cowplot::save_plot
```

#### **Examples**

```
ggplot_save_function(gobject)
```

giotto-class S4 giotto Class

## **Description**

Framework of giotto object to store and work with spatial expression data

#### **Slots**

```
raw_exprs raw expression counts

norm_expr normalized expression counts

norm_scaled_expr normalized and scaled expression counts

custom_expr custom normalized counts

spatial_locs spatial location coordinates for cells

cell_metadata metadata for cells

gene_metadata metadata for genes

cell_ID unique cell IDs

gene_ID unique gene IDs

spatial_network spatial network in data.table/data.frame format

spatial_grid spatial grid in data.table/data.frame format

dimension_reduction slot to save dimension reduction coordinates
```

```
nn_network nearest neighbor network in igraph format parameters slot to save parameters that have been used instructions slot for global function instructions offset_file offset file used to stitch together image fields OS_platform Operating System to run Giotto analysis on
```

heatmSpatialCorGenes heatmSpatialCorGenes

#### **Description**

Create heatmap of spatially correlated genes

### Usage

```
heatmSpatialCorGenes(
  gobject,
  spatCorObject,
  use_clus_name = NULL,
  show_cluster_annot = TRUE,
  show_row_dend = T,
  show_column_dend = F,
  show_row_names = F,
  show_column_names = F,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "heatmSpatialCorGenes",
  ...
)
```

```
gobject
                 giotto object
                 spatial correlation object
spatCorObject
                 name of clusters to visualize (from clusterSpatialCorGenes())
use_clus_name
show_cluster_annot
                 show cluster annotation on top of heatmap
show_row_dend
                 show row dendrogram
show_column_dend
                 show column dendrogram
show_row_names show row names
show_column_names
                 show column names
show_plot
                 show plot
return_plot
                 return ggplot object
```

hyperGeometricEnrich 139

#### Value

Heatmap generated by ComplexHeatmap

### **Examples**

```
heatmSpatialCorGenes(gobject)
```

hyperGeometricEnrich hyperGeometricEnrich

### **Description**

Function to calculate gene signature enrichment scores per spatial position using a hypergeometric test.

#### Usage

```
hyperGeometricEnrich(
  gobject,
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  output_enrichment = c("original", "zscore")
)
```

#### **Arguments**

### **Details**

The enrichment score is calculated based on the p-value from the hypergeometric test, -log10(p-value).

140 loadHMRF

#### Value

data.table with enrichment results

### **Examples**

hyperGeometricEnrich(gobject)

kmeans\_binarize

kmeans\_binarize

## Description

create binarized scores using kmeans

#### Usage

```
kmeans_binarize(x, nstart = 3, iter.max = 10)
```

loadHMRF

loadHMRF

## Description

load previous HMRF

### Usage

```
loadHMRF(
  name_used = "test",
  output_folder_used,
  k_used = 10,
  betas_used,
  python_path_used
)
```

# Arguments

```
name_used name of HMRF that was run
output_folder_used
output folder that was used
k_used number of HMRF domains that was tested
betas_used betas that were tested
python_path_used
python path that was used
```

### **Details**

Description of HMRF parameters ...

makeSignMatrixPAGE 141

#### Value

reloads a previous ran HMRF from doHRMF

#### **Examples**

loadHMRF(gobject)

make Sign Matrix PAGE

make Sign Matrix PAGE

## Description

Function to convert a list of signature genes (e.g. for cell types or processes) into a binary matrix format that can be used with the PAGE enrichment option.

## Usage

```
makeSignMatrixPAGE(sign_names, sign_list)
```

### **Arguments**

sign\_names vector with names for each provided gene signature

sign\_list list of genes (signature)

#### Value

matrix

## See Also

**PAGEEnrich** 

#### **Examples**

makeSignMatrixPAGE()

 ${\tt make Sign Matrix Rank}$ 

makeSignMatrixRank

## Description

Function to convert a single-cell count matrix and a corresponding single-cell cluster vector into a rank matrix that can be used with the Rank enrichment option.

### Usage

```
makeSignMatrixRank(sc_matrix, sc_cluster_ids, gobject = NULL)
```

mergeClusters mergeClusters

### **Arguments**

```
sign_names vector with names for each provided gene signature sign_list list of genes (signature)
```

#### Value

matrix

### See Also

rankEnrich

## **Examples**

```
makeSignMatrixRank()
```

```
make_simulated_network
```

make\_simulated\_network

## Description

Simulate random network.

## Usage

```
make_simulated_network(
  gobject,
  spatial_network_name = "spatial_network",
  cluster_column,
  number_of_simulations = 100
)
```

## **Examples**

```
make_simulated_network(gobject)
```

mergeClusters

mergeClusters

# Description

Merge selected clusters based on pairwise correlation scores and size of cluster.

mergeClusters 143

#### Usage

```
mergeClusters(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  cor = c("pearson", "spearman"),
  new_cluster_name = "merged_cluster",
  min_cor_score = 0.8,
  max_group_size = 20,
  force_min_group_size = 10,
  return_gobject = TRUE,
  verbose = TRUE
)
```

### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
cluster_column name of column to use for clusters
                  correlation score to calculate distance
new_cluster_name
                  new name for merged clusters
                 min correlation score to merge pairwise clusters
min_cor_score
max_group_size max cluster size that can be merged
force_min_group_size
                  size of clusters that will be merged with their most similar neighbor(s)
return_gobject return giotto object
                  be verbose
verbose
```

## Details

Merge selected clusters based on pairwise correlation scores and size of cluster. To avoid large clusters to merge the max\_group\_size can be lowered. Small clusters can be forcibly merged with their most similar pairwise cluster by adjusting the force\_min\_group\_size parameter. Clusters smaller than this value will be merged independent on the provided min\_cor\_score value.

A giotto object is returned by default, if FALSE then the merging vector will be returned.

### Value

Giotto object

```
mergeClusters(gobject)
```

node\_clusters

mygini\_fun

mygini\_fun

## Description

calculate gini coefficient

# Usage

```
mygini_fun(x, weights = rep(1, length(x)))
```

#### Value

gini coefficient

nnDT\_to\_kNN

nnDT\_to\_kNN

## Description

Convert a nearest network data.table to a kNN object

## Usage

```
nnDT_to_kNN(nnDT)
```

### **Arguments**

nnDT

nearest neighbor network in data.table format

### Value

kNN object

 ${\tt node\_clusters}$ 

 $node\_clusters$ 

## Description

Merge selected clusters based on pairwise correlation scores and size of cluster.

## Usage

```
node_clusters(hclus_obj, verbose = TRUE)
```

## Arguments

hclus\_obj hclus object verbose be verbose normalizeGiotto 145

#### Value

list of splitted dendrogram nodes from high to low node height

### **Examples**

```
node_clusters(hclus_obj)
```

normalize Giotto

normalize Giotto

#### **Description**

normalize and/or scale expresion values of Giotto object

## Usage

```
normalizeGiotto(
  gobject,
  norm_methods = c("standard", "osmFISH"),
  library_size_norm = TRUE,
  scalefactor = 6000,
  log_norm = TRUE,
  logbase = 2,
  scale_genes = T,
  scale_cells = T,
  scale_order = c("first_genes", "first_cells"),
  verbose = F
)
```

```
gobject
                  giotto object
norm\_methods
                  normalization method to use
library_size_norm
                  normalize cells by library size
                  scale factor to use after library size normalization
scalefactor
log_norm
                  transform values to log-scale
                  log base to use to log normalize expression values
logbase
scale_genes
                  z-score genes over all cells
scale_cells
                  z-score cells over all genes
scale_order
                  order to scale genes and cells
verbose
                  be verbose
```

OR\_function2

#### **Details**

Currently there are two 'methods' to normalize your raw counts data.

A. The standard method follows the standard protocol which can be adjusted using the provided parameters and follows the following order:

- 1. Data normalization for total library size and scaling by a custom scale-factor.
- 2. Log transformation of data.
- 3. Z-scoring of data by genes and/or cells.
- B. The normalization method as provided by the osmFISH paper is also implemented:
  - 1. First normalize genes, for each gene divide the counts by the total gene count and multiply by the total number of genes.
  - 2. Next normalize cells, for each cell divide the normalized gene counts by the total counts per cell and multiply by the total number of cells.

This data will be saved in the Giotto slot for custom expression.

#### Value

giotto object

### **Examples**

normalizeGiotto(gobject)

OR\_function2

OR\_function2

# Description

calculate odds-ratio

#### Usage

```
OR_function2(A, B, C, D)
```

PAGEEnrich 147

**PAGEEnrich** 

**PAGEEnrich** 

### **Description**

Function to calculate gene signature enrichment scores per spatial position using PAGE.

#### Usage

```
PAGEEnrich(
  gobject,
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  output_enrichment = c("original", "zscore")
)
```

### **Arguments**

```
gobject Giotto object

sign_matrix Matrix of signature genes for each cell type / process
expression_values

expression values to use

reverse_log_scale

reverse expression values from log scale

logbase log base to use if reverse_log_scale = TRUE

output_enrichment

how to return enrichment output
```

### **Details**

sign\_matrix: a binary matrix with genes as row names and cell-types as column names. Alternatively a list of signature genes can be provided to makeSignMatrixPAGE, which will create the matrix for you.

The enrichment Z score is calculated by using method (PAGE) from Kim SY et al., BMC bioinformatics, 2005 as  $Z=((Sm\ mu)*m^(1/2))/delta$ . For each gene in each spot, mu is the fold change values versus the mean expression and delta is the standard deviation. Sm is the mean fold change value of a specific marker gene set and m is the size of a given marker gene set.

#### Value

data.table with enrichment results

### See Also

makeSignMatrixPAGE

```
PAGEEnrich(gobject)
```

plotCPGscores

pDataDT

pDataDT

## Description

show cell metadata

## Usage

```
pDataDT(gobject)
```

## Arguments

gobject

giotto object

### Value

data.table with cell metadata

## **Examples**

```
pDataDT(gobject)
```

plotCPGscores

plotCPGscores

# Description

Create heatmap from cell-cell proximity scores

# Usage

```
plotCPGscores(
   CPGscores,
   selected_interactions = NULL,
   selected_genes = NULL,
   detail_plot = T,
   simple_plot = F,
   simple_plot_facet = c("interaction", "genes"),
   facet_scales = "fixed",
   facet_ncol = length(selected_genes),
   facet_nrow = length(selected_interactions),
   show_plot = F
)
```

plotGTGscores 149

### **Arguments**

```
CPGscores
                  CPGscores, output from getCellProximityGeneScores()
selected_interactions
                  interactions to show
selected_genes genes to show
detail_plot
                  show detailed info in both interacting cell types
simple_plot
                  show a simplified plot
simple_plot_facet
                  facet on interactions or genes with simple plot
                  ggplot facet scales paramter
facet_scales
facet_ncol
                  ggplot facet ncol parameter
facet_nrow
                  ggplot facet nrow parameter
show_plot
                  show plot
```

#### **Details**

Give more details ...

#### Value

ggplot barplot

#### **Examples**

```
plotCPGscores(CPGscores)
```

 ${\tt plotGTGscores}$ 

plotGTGscores

### **Description**

Create heatmap from cell-cell proximity scores

## Usage

```
plotGTGscores(
   gobject,
   GTGscore,
   selected_interactions = NULL,
   selected_gene_to_gene = NULL,
   detail_plot = T,
   simple_plot = F,
   simple_plot_facet = c("interaction", "genes"),
   facet_scales = "fixed",
   facet_ncol = length(selected_gene_to_gene),
   facet_nrow = length(selected_interactions),
   colors = c("blue", "red"),
   show_plot = NA,
   return_plot = NA,
```

150 plotGTGscores

```
save_plot = NA,
save_param = list(),
default_save_name = "plotGTGscores"
)
```

## Arguments

gobject giotto object

GTGscore GTGscore, output from getGeneToGeneScores()

selected\_interactions

interactions to show

detail\_plot show detailed info in both interacting cell types

simple\_plot show a simplified plot

simple\_plot\_facet

facet on interactions or genes with simple plot

facet\_scales ggplot facet scales paramter
facet\_ncol ggplot facet ncol parameter
facet\_nrow ggplot facet nrow parameter

colors vector with 2 colors to represent respectively all and selected cells

show\_plot show plots

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

selected\_genes genes to show

### **Details**

Give more details ...

## Value

ggplot barplot

```
plotGTGscores(GTGscore)
```

plotHeatmap 151

plotHeatmap

plotHeatmap

### **Description**

Creates heatmap for genes and clusters.

### Usage

```
plotHeatmap(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  cluster_column = NULL,
  cluster_order = c("size", "correlation", "custom"),
  cluster_custom_order = NULL,
  cluster_color_code = NULL,
  cluster_cor_method = "pearson",
  cluster_hclust_method = "ward.D"
  gene_order = c("custom", "correlation"),
  gene_custom_order = NULL,
  gene_cor_method = "pearson",
  gene_hclust_method = "complete",
  show_values = c("rescaled", "z-scaled", "original"),
  size_vertical_lines = 1.1,
  gradient_colors = c("blue", "yellow", "red"),
  gene_label_selection = NULL,
  axis_text_y_size = NULL,
  legend_nrows = 1,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotHeatmap"
)
```

```
gobject giotto object
expression_values
expression values to use
genes genes to use
cluster_column name of column to use for clusters
cluster_order method to determine cluster order
cluster_custom_order
custom order for clusters
cluster_color_code
color code for clusters
cluster_cor_method
method for cluster correlation
```

152 plotHeatmap

```
cluster_hclust_method
                 method for hierarchical clustering of clusters
gene_order
                 method to determine gene order
gene_custom_order
                 custom order for genes
gene_cor_method
                 method for gene correlation
gene_hclust_method
                 method for hierarchical clustering of genes
show_values
                 which values to show on heatmap
size_vertical_lines
                 sizes for vertical lines
gradient_colors
                 colors for heatmap gradient
gene_label_selection
                 subset of genes to show on y-axis
axis_text_y_size
                 size for y-axis text
legend_nrows
                 number of rows for the cluster legend
show_plot
                 show plot
return_plot
                 return ggplot object
                 directly save the plot [boolean]
save_plot
                 list of saving parameters from all_plots_save_function
save_param
default_save_name
                 default save name
```

### **Details**

If you want to display many genes there are 2 ways to proceed:

- 1. set axis\_text\_y\_size to a really small value and show all genes
- 2. provide a subset of genes to display to gene\_label\_selection

### Value

ggplot

### **Examples**

plotHeatmap(gobject)

plotly\_axis\_scale\_2D 153

```
plotly_axis_scale_2D plotly_axis_scale_2D
```

# Description

adjust the axis scale in 3D plotly plot

### Usage

```
plotly_axis_scale_2D(
  cell_locations,
  sdimx = NULL,
  sdimy = NULL,
  mode = c("cube", "real", "custom"),
  custom_ratio = NULL
)
```

#### **Arguments**

```
cell_locations spatial_loc in giotto object sdimx x axis of cell spatial location sdimy y axis of cell spatial location mode axis adjustment mode custom_ratio set the ratio artificially
```

#### Value

edges in spatial grid as data.table()

### **Examples**

```
plotly_axis_scale_2D(gobject)
```

```
plotly_axis_scale_3D plotly_axis_scale_3D
```

## Description

adjust the axis scale in 3D plotly plot

## Usage

```
plotly_axis_scale_3D(
  cell_locations,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  mode = c("cube", "real", "custom"),
  custom_ratio = NULL
)
```

154 plotly\_grid

### **Arguments**

#### Value

edges in spatial grid as data.table()

### **Examples**

```
plotly_axis_scale_3D(gobject)
```

plotly\_grid

plotly\_grid

## Description

provide grid segment to draw in plot\_ly()

### Usage

```
plotly_grid(
   spatial_grid,
   x_start = "x_start",
   y_start = "y_start",
   x_end = "x_end",
   y_end = "y_end"
)
```

# Arguments

```
spatial_grid spatial_grid in giotto object
```

### Value

edges in spatial grid as data.table()

```
plotly_grid(gobject)
```

plotly\_network 155

plotly\_network

plotly\_network

### **Description**

provide network segment to draw in 3D plot\_ly()

### Usage

```
plotly_network(
  network,
  x = "sdimx_begin",
  y = "sdimy_begin",
  z = "sdimz_begin",
  x_end = "sdimx_end",
  y_end = "sdimy_end",
  z_end = "sdimz_end"
)
```

### **Arguments**

gobject network in giotto object

#### Value

edges in network as data.table()

### **Examples**

```
plotly_network(gobject)
```

```
plotMetaDataCellsHeatmap
```

plot Meta Data Cells Heat map

### **Description**

Creates heatmap for numeric cell metadata within aggregated clusters.

### Usage

```
plotMetaDataCellsHeatmap(
  gobject,
  metadata_cols = NULL,
  spat_enr_names = NULL,
  value_cols = NULL,
  first_meta_col = NULL,
  second_meta_col = NULL,
  show_values = c("zscores", "original", "zscores_rescaled"),
  custom_cluster_order = NULL,
```

custom\_gene\_order

```
clus_cor_method = "pearson",
      clus_cluster_method = "complete",
      custom_values_order = NULL,
      values_cor_method = "pearson",
      values_cluster_method = "complete",
      midpoint = 0,
      x_{text_size} = 8,
      x_{text_angle} = 45,
      y_text_size = 8,
      strip_text_size = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "plotMetaDataCellsHeatmap"
    )
Arguments
    gobject
                     giotto object
                     annotation columns found in pDataDT(gobject)
    metadata_cols
    spat_enr_names spatial enrichment results to include
    value_cols
                     value columns to use
    first_meta_col if more than 1 metadata column, select the x-axis factor
    second_meta_col
                     if more than 1 metadata column, select the facetting factor
    show_values
                     which values to show on heatmap
    custom_cluster_order
                     custom cluster order (default = NULL)
    clus_cor_method
                     correlation method for clusters
    clus_cluster_method
                     hierarchical cluster method for the clusters
                     midpoint of show_values
    midpoint
                     size of x-axis text
    x_text_size
    x_text_angle
                     angle of x-axis text
    y_text_size
                     size of y-axis text
    strip_text_size
                     size of strip text
    show_plot
                     show plot
    return_plot
                     return ggplot object
    save_plot
                     directly save the plot [boolean]
                     list of saving parameters from all_plots_save_function
    save_param
    default_save_name
                     default save name for saving, don't change, change save_name in save_param
```

custom gene order (default = NULL)

plotMetaDataHeatmap 157

#### **Details**

Creates heatmap for the average values of selected value columns in the different annotation groups.

### Value

ggplot or data.table

#### See Also

plotMetaDataHeatmap for gene expression instead of numeric cell annotation data.

#### **Examples**

```
plotMetaDataCellsHeatmap(gobject)
```

plotMetaDataHeatmap

### **Description**

Creates heatmap for genes within aggregated clusters.

### Usage

```
plotMetaDataHeatmap(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  metadata_cols = NULL,
  selected_genes = NULL,
  first_meta_col = NULL,
  second_meta_col = NULL,
  show_values = c("zscores", "original", "zscores_rescaled"),
  custom_cluster_order = NULL,
  clus_cor_method = "pearson",
  clus_cluster_method = "complete",
  custom_gene_order = NULL,
  gene_cor_method = "pearson",
  gene_cluster_method = "complete",
  midpoint = 0,
  x_{text_size} = 10,
  x_{text_angle} = 45,
  y_{text_size} = 10,
  strip_text_size = 8,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
```

```
save_param = list(),
default_save_name = "plotMetaDataHeatmap"
)
```

#### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
metadata_cols annotation columns found in pDataDT(gobject)
selected_genes subset of genes to use
first_meta_col if more than 1 metadata column, select the x-axis factor
second_meta_col
                  if more than 1 metadata column, select the facetting factor
                  which values to show on heatmap
show_values
custom_cluster_order
                  custom cluster order (default = NULL)
clus_cor_method
                  correlation method for clusters
clus_cluster_method
                  hierarchical cluster method for the clusters
custom_gene_order
                  custom gene order (default = NULL)
gene_cor_method
                  correlation method for genes
gene_cluster_method
                  hierarchical cluster method for the genes
midpoint
                  midpoint of show_values
x_text_size
                  size of x-axis text
                  angle of x-axis text
x_text_angle
y_text_size
                  size of y-axis text
strip_text_size
                  size of strip text
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name
```

# **Details**

Creates heatmap for the average expression of selected genes in the different annotation/cluster groups

### Value

ggplot or data.table

plotPCA 159

#### See Also

plotMetaDataCellsHeatmap for numeric cell annotation instead of gene expression.

#### **Examples**

```
plotMetaDataHeatmap(gobject)
```

plotPCA

plotPCA

### **Description**

Short wrapper for PCA visualization

#### Usage

```
plotPCA(gobject, dim_reduction_name = "pca", default_save_name = "PCA", ...)
```

```
gobject
                 giotto object
dim_reduction_name
                 dimension reduction name
default_save_name
                 default save name for saving, don't change, change save_name in save_param
groub_by
                 create multiple plots based on cell annotation column
group_by_subset
                 subset the group_by factor column
dim1_to_use
                 dimension to use on x-axis
                 dimension to use on y-axis
dim2_to_use
spat_enr_names names of spatial enrichment results to include
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
                 name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                 color for cells (see details)
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                 midpoint for color gradient
gradient_limits
                 vector with lower and upper limits
```

160 plotPCA

```
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
                  size of labels
label_size
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
title
                  title for plot, defaults to cell_color parameter
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

#### **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotPCA\_3D

### Value

ggplot

## **Examples**

plotPCA(gobject)

plotPCA\_2D 161

plotPCA\_2D plotPCA\_2D

#### **Description**

Short wrapper for PCA visualization

## Usage

```
plotPCA_2D(
  gobject,
  dim_reduction_name = "pca",
  default_save_name = "PCA_2D",
)
```

```
gobject
                 giotto object
dim_reduction_name
                 dimension reduction name
default_save_name
                 default save name for saving, don't change, change save_name in save_param
                 create multiple plots based on cell annotation column
groub_by
group_by_subset
                 subset the group_by factor column
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
network_name
                 name of NN network to use, if show_NN_network = TRUE
                 color for cells (see details)
cell_color
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                 select subset of cells/clusters based on cell_color parameter
```

plotPCA\_2D

```
select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
title
                  title for plot, defaults to cell_color parameter
show_legend
                  show legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
                  cowplot param: relative width
cow_rel_w
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
```

## **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotPCA\_3D

#### Value

ggplot

```
plotPCA_2D(gobject)
```

plotPCA\_3D 163

plotPCA\_3D plotPCA\_3D

#### **Description**

Visualize cells according to 3D PCA dimension reduction

# Usage

```
plotPCA_3D(
  gobject,
  dim_reduction_name = "pca",
  default_save_name = "PCA_3D",
   ...
)
```

```
gobject
                  giotto object
dim_reduction_name
                  pca dimension reduction name
default_save_name
                  default save name for saving, ideally change save_name in save_param
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
dim3_to_use
                  dimension to use on z-axis
show_NN_network
                  show underlying NN network
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
```

164 plotTSNE

```
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
                  size of labels
label_size
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
show_legend
                  show legend
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

#### **Details**

Description of parameters.

#### Value

plotly

# **Examples**

```
plotPCA_3D(gobject)
```

plotTSNE plotTSNE

### **Description**

Short wrapper for tSNE visualization

## Usage

```
plotTSNE(gobject, dim_reduction_name = "tsne", default_save_name = "tSNE", ...)
```

plotTSNE 165

dim2\_to\_use dimension to use on y-axis spat\_enr\_names names of spatial enrichment results to include show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) network\_name name of NN network to use, if show\_NN\_network = TRUE cell\_color color for cells (see details) color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_size size of point (cell) point\_border\_col color of border around points point\_border\_stroke stroke size of border around points title title for plot, defaults to cell\_color parameter show\_legend show legend legend\_text size of legend text axis\_text size of axis text axis\_title size of axis title

166 plotTSNE\_2D

```
cow_n_col
                 cowplot param: how many columns
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
cow_align
                 cowplot param: how to align
show_plot
                 show plot
                 return ggplot object
return_plot
save_plot
                 directly save the plot [boolean]
                 list of saving parameters from all_plots_save_function
save_param
```

### **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotTSNE\_3D

#### Value

ggplot

### **Examples**

```
plotTSNE(gobject)
```

plotTSNE\_2D

plotTSNE\_2D

### **Description**

Short wrapper for tSNE visualization

### Usage

```
plotTSNE_2D(
  gobject,
  dim_reduction_name = "tsne",
  default_save_name = "tSNE_2D",
  ...
)
```

plotTSNE\_2D 167

dim2\_to\_use dimension to use on y-axis spat\_enr\_names names of spatial enrichment results to include show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) network\_name name of NN network to use, if show\_NN\_network = TRUE cell\_color color for cells (see details) color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_size size of point (cell) point\_border\_col color of border around points point\_border\_stroke stroke size of border around points title title for plot, defaults to cell\_color parameter show\_legend show legend size of legend text legend\_text axis\_text size of axis text axis\_title size of axis title

168 plotTSNE\_3D

```
cow_n_col
                 cowplot param: how many columns
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
                 cowplot param: how to align
cow_align
show_plot
                 show plot
                 return ggplot object
return_plot
                 directly save the plot [boolean]
save_plot
                 list of saving parameters from all_plots_save_function
save_param
```

### **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotTSNE\_3D

#### Value

ggplot

### **Examples**

```
plotTSNE_2D(gobject)
```

plotTSNE\_3D plotTSNE\_3D

### **Description**

Visualize cells according to dimension reduction coordinates

### Usage

```
plotTSNE_3D(
  gobject,
  dim_reduction_name = "tsne",
  default_save_name = "TSNE_3D",
  ...
)
```

plotTSNE\_3D

```
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
edge_alpha
                  column to use for alpha of the edges
                  size of point (cell)
point_size
show_legend
                  show legend
                  show plot
show_plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

### **Details**

Description of parameters.

### Value

plotly

```
plotTSNE_3D(gobject)
```

170 plotUMAP

plotUMAP plotUMAP

### **Description**

Short wrapper for UMAP visualization

### Usage

```
plotUMAP(gobject, dim_reduction_name = "umap", default_save_name = "UMAP", ...)
```

```
gobject
                  giotto object
dim_reduction_name
                  dimension reduction name
default_save_name
                  default save name for saving, don't change, change save_name in save_param
groub_by
                  create multiple plots based on cell annotation column
group_by_subset
                  subset the group_by factor column
                  dimension to use on x-axis
dim1_to_use
                  dimension to use on y-axis
dim2_to_use
spat_enr_names names of spatial enrichment results to include
show_NN_network
                  show underlying NN network
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
network_name
                  name of NN network to use, if show_NN_network = TRUE
                  color for cells (see details)
cell_color
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
```

plotUMAP 171

```
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
title
                  title for plot, defaults to cell_color parameter
show_legend
                  show legend
legend_text
                  size of legend text
axis_text
                  size of axis text
                  size of axis title
axis_title
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
                  cowplot param: relative width
cow_rel_w
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

### **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotUMAP\_3D

## Value

ggplot

```
plotUMAP(gobject)
```

172 plotUMAP\_2D

plotUMAP\_2D

plotUMAP\_2D

### **Description**

Short wrapper for UMAP visualization

## Usage

```
plotUMAP_2D(
  gobject,
  dim_reduction_name = "umap",
  default_save_name = "UMAP_2D",
)
```

```
gobject
                 giotto object
dim_reduction_name
                 dimension reduction name
default_save_name
                 default save name for saving, don't change, change save_name in save_param
                 create multiple plots based on cell annotation column
groub_by
group_by_subset
                 subset the group_by factor column
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
network_name
                 name of NN network to use, if show_NN_network = TRUE
                 color for cells (see details)
cell_color
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                 select subset of cells/clusters based on cell_color parameter
```

plotUMAP\_2D 173

```
select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
title
                  title for plot, defaults to cell_color parameter
show_legend
                  show legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
```

### **Details**

Description of parameters, see dimPlot2D. For 3D plots see plotUMAP\_3D

#### Value

ggplot

```
plotUMAP_2D(gobject)
```

174 plotUMAP\_3D

plotUMAP\_3D

plotUMAP\_3D

#### **Description**

Visualize cells according to dimension reduction coordinates

## Usage

```
plotUMAP_3D(
  gobject,
  dim_reduction_name = "umap",
  default_save_name = "UMAP_3D",
  ...
)
```

```
gobject
                  giotto object
dim_reduction_name
                  umap dimension reduction name
default_save_name
                  default save name for saving, don't change, change save_name in save_param
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
dim3_to_use
                  dimension to use on z-axis
show_NN_network
                  show underlying NN network
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
```

```
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
                  column to use for alpha of the edges
edge_alpha
point_size
                  size of point (cell)
show_legend
                  show legend
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

#### **Details**

Description of parameters.

### Value

plotly

#### **Examples**

```
plotUMAP_3D(gobject)
```

### **Description**

Visualize cells in network layer according to dimension reduction coordinates

# Usage

```
plot_network_layer_ggplot(
   ggobject,
   annotated_network_DT,
   edge_alpha = NULL,
   show_legend = T
)
```

### **Arguments**

```
annotated_network_DT
```

annotated network data.table of selected cells

edge\_alpha alpha of network edges

show\_legend show legend gobject giotto object

#### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
plot_network_layer_ggplot(gobject)
```

### **Description**

Visualize cells in point layer according to dimension reduction coordinates

### Usage

```
plot_point_layer_ggplot(
  ggobject,
  annotated_DT_selected,
  annotated_DT_other,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = 0,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_legend = T
)
```

#### **Arguments**

```
annotated_DT_selected
                  annotated data.table of selected cells
annotated_DT_other
                  annotated data.table of not selected cells
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
                  show legend
show_legend
                  giotto object
gobject
```

### **Details**

Description of parameters.

### Value

ggplot

#### **Examples**

```
plot_point_layer_ggplot(gobject)
```

### **Description**

creat ggplot point layer for spatial coordinates

### Usage

```
plot_spat_point_layer_ggplot(
  ggobject,
  sdimx = NULL,
  sdimy = NULL,
  cell_locations_metadata_selected,
  cell_locations_metadata_other,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  point_size = 2,
  point_border_col = "lightgrey",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 1,
  show_legend = TRUE
)
```

```
sdimx x-axis dimension name (default = 'sdimx')
sdimy y-axis dimension name (default = 'sdimy')
cell_locations_metadata_selected
annotated location from selected cells
```

```
cell_locations_metadata_other
                  annotated location from non-selected cells
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
                  size of labels
label_size
label_fontface font of labels
show_other_cells
                  display not selected cells
other_cell_color
                  color for not selected cells
other_point_size
                  point size for not selected cells
show_legend
                  show legend
                  giotto object
gobject
```

## **Details**

Description of parameters.

### Value

ggplot

```
plot_spat_point_layer_ggplot(gobject)
```

180 rankEnrich

print.giotto

print method for giotto class

#### **Description**

print method for giotto class. Prints the chosen number of genes (rows) and cells (columns) from the raw count matrix. Also print the spatial locations for the chosen number of cells.

### Usage

```
print.giotto(object, ...)
```

### **Arguments**

nr\_genes number of genes (rows) to print nr\_cells number of cells (columns) to print

rankEnrich

rankEnrich

## Description

Function to calculate gene signature enrichment scores per spatial position using a rank based approach.

### Usage

```
rankEnrich(
  gobject,
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  output_enrichment = c("original", "zscore")
)
```

rankSpatialCorGroups 181

#### **Details**

sign\_matrix: a rank-fold matrix with genes as row names and cell-types as column names. Alternatively a scRNA-seq matrix and vector with clusters can be provided to makeSignMatrixRank, which will create the matrix for you.

First a new rank is calculated as  $R = (R1*R2)^{\Lambda}(1/2)$ , where R1 is the rank of fold-change for each gene in each spot and R2 is the rank of each marker in each cell type. The Rank-Biased Precision is then calculated as:  $RBP = (1 - 0.99) * (0.99)^{\Lambda}(R - 1)$  and the final enrichment score is then calculated as the sum of top 100 RBPs.

### Value

data.table with enrichment results

#### See Also

```
makeSignMatrixRank
```

### **Examples**

```
rankEnrich(gobject)
```

```
rankSpatialCorGroups rankSpatialCorGroups
```

## Description

Rank spatial correlated clusters according to correlation structure

## Usage

```
rankSpatialCorGroups(
  gobject,
  spatCorObject,
  use_clus_name = NULL,
  show_plot = NA,
  return_plot = FALSE,
  save_plot = NA,
  save_param = list(),
  default_save_name = "rankSpatialCorGroups"
)
```

### **Arguments**

```
gobject giotto object
spatCorObject spatial correlation object
use_clus_name name of clusters to visualize (from clusterSpatialCorGenes())
show_plot show plot
return_plot return ggplot object
```

182 readGiottoInstructions

### Value

data.table with positive (within group) and negative (outside group) scores

### **Examples**

```
rankSpatialCorGroups(gobject)
```

rank\_binarize

rank\_binarize

## **Description**

create binarized scores using arbitrary rank of top genes

## Usage

```
rank\_binarize(x, max\_rank = 200)
```

readGiottoInstructions

 ${\it readGiottoInstrunctions}$ 

## Description

Retrieves the instruction associated with the provided parameter

## Usage

```
readGiottoInstructions(giotto_instructions, param = NULL)
```

### **Arguments**

```
giotto_instructions
```

giotto object or result from createGiottoInstructions()

param parameter to retrieve

### Value

specific parameter

## **Examples**

readGiottoInstrunctions()

removeCellAnnotation 183

```
remove Cell Annotation remove Cell Annotation
```

## Description

removes cell annotation of giotto object

### Usage

```
removeCellAnnotation(gobject, columns = NULL, return_gobject = TRUE)
```

### **Arguments**

gobject giotto object

columns names of columns to remove

return\_gobject boolean: return giotto object (default = TRUE)

### **Details**

if return\_gobject = FALSE, it will return the cell metadata

### Value

giotto object

## **Examples**

removeCellAnnotation(gobject)

removeGeneAnnotation removeGeneAnnotation

## **Description**

removes gene annotation of giotto object

# Usage

```
removeGeneAnnotation(gobject, columns = NULL, return_gobject = TRUE)
```

### **Arguments**

gobject giotto object

columns names of columns to remove

return\_gobject boolean: return giotto object (default = TRUE)

### **Details**

if return\_gobject = FALSE, it will return the gene metadata

runPCA

### Value

```
giotto object
```

### **Examples**

```
removeGeneAnnotation(gobject)
```

```
replaceGiottoInstructions
```

replace Giot to Instructions

## Description

Function to replace all instructions from giotto object

## Usage

```
replaceGiottoInstructions(gobject, instructions = NULL)
```

### **Arguments**

```
gobject giotto object
instructions new instructions (e.g. result from createGiottoInstructions)
```

### Value

named vector with giotto instructions

### **Examples**

```
replaceGiottoInstructions()
```

runPCA

runPCA

## Description

runs a Principal Component Analysis

```
runPCA(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  name = "pca",
  genes_to_use = NULL,
  return_gobject = TRUE,
  scale_unit = F,
  ncp = 200,
  ...
)
```

runtSNE 185

### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
                  cells or genes
reduction
name
                  arbitrary name for PCA run
                  subset of genes to use for PCA
genes_to_use
return_gobject boolean: return giotto object (default = TRUE)
scale_unit
                  scale features before PCA
                  number of principal components to calculate
ncp
                  additional parameters for PCA (see details)
. . .
```

### **Details**

See PCA for more information about other parameters.

### Value

giotto object with updated PCA dimension recuction

### **Examples**

```
runPCA(gobject)
```

runtSNE

runtSNE

## Description

run tSNE

```
runtSNE(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  dim_reduction_to_use = "pca",
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  name = "tsne",
  genes_to_use = NULL,
  return_gobject = TRUE,
  dims = 2,
  perplexity = 30,
  theta = 0.5,
  do_PCA_first = F,
  set\_seed = T,
  seed_number = 1234,
)
```

186 runtSNE

### **Arguments**

gobject giotto object

expression\_values

expression values to use

reduction cells or genes

dim\_reduction\_to\_use

use another dimension reduction set as input

dim\_reduction\_name

name of dimension reduction set to use

dimensions\_to\_use

number of dimensions to use as input

name arbitrary name for tSNE run

genes\_to\_use if dim\_reduction\_to\_use = NULL, which genes to use

return\_gobject boolean: return giotto object (default = TRUE)

dims tSNE param: number of dimensions to return

perplexity tSNE param: perplexity

theta tSNE param: theta

do\_PCA\_first tSNE param: do PCA before tSNE (default = FALSE)

set\_seed use of seed

seed\_number seed number to use

... additional tSNE parameters

### **Details**

See Rtsne for more information about these and other parameters.

- Input for tSNE dimension reduction can be another dimension reduction (default = 'pca')
- To use gene expression as input set dim\_reduction\_to\_use = NULL
- multiple tSNE results can be stored by changing the *name* of the analysis

### Value

giotto object with updated tSNE dimension recuction

### **Examples**

runtSNE(gobject)

runUMAP 187

runUMAP runUMAP

## **Description**

run UMAP

### Usage

```
runUMAP(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  dim_reduction_to_use = "pca",
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  name = "umap",
  genes_to_use = NULL,
  return_gobject = TRUE,
  n_neighbors = 40,
  n_{components} = 2,
  n_{epochs} = 400,
  min_dist = 0.01,
  n_{threads} = 1,
  spread = 5,
  set\_seed = T,
  seed_number = 1234,
)
```

## **Arguments**

```
gobject
                 giotto object
expression_values
                 expression values to use
                 cells or genes
reduction
dim_reduction_to_use
                 use another dimension reduction set as input
dim_reduction_name
                 name of dimension reduction set to use
dimensions_to_use
                 number of dimensions to use as input
                 arbitrary name for UMAP run
name
                 if dim_reduction_to_use = NULL, which genes to use
genes_to_use
return_gobject boolean: return giotto object (default = TRUE)
n\_neighbors
                 UMAP param: number of neighbors
                 UMAP param: number of components
n_components
n_epochs
                 UMAP param: number of epochs
```

188 selectPatternGenes

```
min_dist UMAP param: minimum distance
n_threads UMAP param: threads to use
spread UMAP param: spread
set_seed use of seed
seed_number seed number to use
additional UMAP parameters
```

### **Details**

See umap for more information about these and other parameters.

- Input for UMAP dimension reduction can be another dimension reduction (default = 'pca')
- To use gene expression as input set dim\_reduction\_to\_use = NULL
- multiple UMAP results can be stored by changing the *name* of the analysis

### Value

giotto object with updated UMAP dimension recuction

#### **Examples**

```
runUMAP(gobject)
```

selectPatternGenes selectPatternGenes

### **Description**

Select genes correlated with spatial patterns

### Usage

```
selectPatternGenes(
   spatPatObj,
   dimensions = 1:5,
   top_pos_genes = 10,
   top_neg_genes = 10,
   min_pos_cor = 0.5,
   min_neg_cor = -0.5,
   return_top_selection = FALSE
)
```

### Arguments

spatPatObj

```
dimensions dimensions to identify correlated genes for.

top_pos_genes Top positively correlated genes.

top_neg_genes Top negatively correlated genes.

min_pos_cor Minimum positive correlation score to include a gene.

min_neg_cor Minimum negative correlation score to include a gene.
```

Output from detectSpatialPatterns

189

### **Details**

Description.

### Value

Data.table with genes associated with selected dimension (PC).

## **Examples**

```
selectPatternGenes(gobject)
```

```
select\_expression\_values
```

select\_expression\_values

# Description

helper function to select expression values

## Usage

```
select_expression_values(gobject, values)
```

## Arguments

gobject giotto object

values expression values to extract

## Value

expression matrix

show,giotto-method

show method for giotto class

## Description

show method for giotto class

```
## S4 method for signature 'giotto'
show(object)
```

 $show Cluster Dendrogram \quad show Cluster Dendrogram \quad$ 

### **Description**

Creates dendrogram for selected clusters.

#### Usage

```
showClusterDendrogram(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  cor = c("pearson", "spearman"),
  distance = "ward.D",
  h = NULL,
  h_color = "red",
  rotate = FALSE,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "showClusterDendrogram",
  ...
)
```

### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
cluster_column name of column to use for clusters
                  correlation score to calculate distance
cor
distance
                  distance method to use for hierarchical clustering
h
                  height of horizontal lines to plot
h_color
                  color of horizontal lines
                  rotate dendrogram 90 degrees
rotate
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  additional parameters for ggdendrogram()
```

### **Details**

Expression correlation dendrogram for selected clusters.

showClusterHeatmap 191

### Value

ggplot

### **Examples**

showClusterDendrogram(gobject)

showClusterHeatmap

*showClusterHeatmap* 

### **Description**

Creates heatmap based on identified clusters

## Usage

```
showClusterHeatmap(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = "all",
  cluster_column,
  cor = c("pearson", "spearman"),
  distance = "ward.D",
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "showClusterHeatmap",
  ...
)
```

# Arguments

```
giotto object
gobject
expression_values
                  expression values to use
                  vector of genes to use, default to 'all'
genes
cluster_column name of column to use for clusters
                  correlation score to calculate distance
cor
distance
                  distance method to use for hierarchical clustering
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  additional parameters for the Heatmap function from ComplexHeatmap
. . .
```

192 showCPGscores

### **Details**

Correlation heatmap of selected clusters.

#### Value

ggplot

### **Examples**

showClusterHeatmap(gobject)

showCPGscores

showCPGscores

## **Description**

visualize Cell Proximity Gene enrichment scores

## Usage

```
showCPGscores(
 gobject,
 CPGscore,
 method = c("volcano", "cell_barplot", "cell-cell", "cell_sankey"),
 min_cells = 5,
 min_fdr = 0.05,
 min_spat_diff = 0.2,
 min_log2_fc = 0.5,
 keep_int_duplicates = TRUE,
 direction = c("both", "up", "down"),
 cell_color_code = NULL,
  show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "showCPGscores"
)
```

## Arguments

```
CPGscore
                 CPGscore, output from getCellProximityGeneScores()
method
                 visualization method
                 min number of cells threshold
min_cells
min_fdr
                 fdr threshold
min_spat_diff
                 spatial difference threshold
                 min log2 fold-change
min_log2_fc
keep_int_duplicates
                 keep both cell_A-cell_B and cell_B-cell_A
direction
                 up or downregulation or both
```

```
cell_color_code
```

color code for cell types

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

### **Details**

Give more details ...

## Value

Gene to gene scores in data.table format

### **Examples**

```
showCPGscores(CPGscore)
```

showGeneExpressionProximityScore

show Gene Expression Proximity Score

## Description

Create heatmap from cell-cell proximity scores

# Usage

```
showGeneExpressionProximityScore(
  scores,
  selected_gene,
  sort_column = "diff_spat"
)
```

# Arguments

scores CPscore, output from getAverageCellProximityGeneScores()

selected\_gene gene to show

sort\_column column name to use for sorting

### **Details**

Give more details ...

### Value

ggplot barplot

showGTGscores

### **Examples**

```
showGeneExpressionProximityScore(scores)
```

```
showGiottoInstructions
```

showGiottoInstructions

## Description

Function to display all instructions from giotto object

## Usage

```
showGiottoInstructions(gobject)
```

## Arguments

```
gobject giotto object
```

### Value

named vector with giotto instructions

### **Examples**

```
showGiottoInstructions()
```

 $\verb|showGTGscores||$ 

show GTG scores

## Description

visualize Cell Proximity Gene enrichment scores

```
showGTGscores(
 GTGscore,
 method = c("cell_barplot", "cell-cell", "cell_sankey"),
 min_cells = 5,
 min_pval = 0.05,
 min_spat_diff = 0.2,
 min_log2_fc = 0.5,
 direction = c("both", "up", "down"),
 cell_color_code = NULL,
 show_plot = T,
  specific_genes_1 = NULL,
 specific_genes_2 = NULL,
 first_cell_name = "ligand cell",
 second_cell_name = "receptor cell",
 return_DT = F
)
```

### **Arguments**

```
visualization method
method
min_cells
                  min number of cells threshold
min_pval
                  p-value threshold
min_spat_diff
                  spatial difference threshold
                  log2 fold-change threshold
min_log2_fc
direction
                  up or downregulation or both
cell_color_code
                  color code for cell types
show_plot
                  print plot
specific_genes_1
                  subset of genes, matched with specific_genes_2
specific_genes_2
                  subset of genes, matched with specific_genes_1
first_cell_name
                  name for first cells
second_cell_name
                  name for second cells
                  CPGscore, output from getCellProximityGeneScores()
CPGscore
```

### **Details**

Give more details ...

### Value

ggplot

### **Examples**

```
showGTGscores(CPGscore)
```

```
show Int {\tt Expression Proximity Score} \\ show Int {\tt Expression Proximity Score}
```

## Description

Create heatmap from cell-cell proximity scores

```
showIntExpressionProximityScore(
   scores,
   selected_interaction,
   sort_column = "diff_spat",
   show_enriched_n = 5,
   show_depleted_n = 5
)
```

196 showPattern

### **Arguments**

scores scores, output from getAverageCellProximityGeneScores()
selected\_interaction
interaction to show
sort\_column column name to use for sorting
show\_enriched\_n
show top enriched interactions

show\_depleted\_n

show top depleted interactions

### **Details**

Give more details ...

#### Value

ggplot barplot

### **Examples**

showIntExpressionProximityScore(scores)

showPattern

showPattern

### **Description**

show patterns for 2D spatial data

## Usage

```
showPattern(gobject, spatPatObj, ...)
```

## **Arguments**

gobject giotto object

spatPatObj Output from detectSpatialPatterns

dimension dimension to plot

trim Trim ends of the PC values.

background\_color

background color for plot

grid\_border\_color

color for grid

show\_legend show legend of ggplot

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

showPattern2D

### Value

ggplot

### See Also

showPattern2D

### **Examples**

```
showPattern(gobject)
```

showPattern2D

showPattern2D

## Description

show patterns for 2D spatial data

# Usage

```
showPattern2D(
  gobject,
  spatPatObj,
  dimension = 1,
  trim = c(0.02, 0.98),
  background_color = "white",
  grid_border_color = "grey",
  show_legend = T,
  point_size = 1,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "showPattern2D"
)
```

# Arguments

return\_plot

giotto object gobject Output from detectSpatialPatterns spatPatObj dimension dimension to plot Trim ends of the PC values. trim background\_color background color for plot grid\_border\_color color for grid show\_legend show legend of ggplot show\_plot show plot

return ggplot object

198 showPattern3D

#### Value

ggplot

## **Examples**

```
showPattern2D(gobject)
```

showPattern3D

showPattern3D

## Description

show patterns for 3D spatial data

## Usage

```
showPattern3D(
 gobject,
 spatPatObj,
 dimension = 1,
  trim = c(0.02, 0.98),
 background_color = "white",
 grid_border_color = "grey",
 show_legend = T,
 point_size = 1,
 axis_scale = c("cube", "real", "custom"),
 custom_ratio = NULL,
 x_{ticks} = NULL,
 y_ticks = NULL,
 z_ticks = NULL,
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "showPattern3D"
)
```

## Arguments

```
gobject giotto object
spatPatObj Output from detectSpatialPatterns
dimension dimension to plot
trim Trim ends of the PC values.
background_color
background color for plot
```

showPatternGenes 199

```
grid_border_color
                  color for grid
show_legend
                  show legend of plot
                  adjust the point size
point_size
axis_scale
                  scale the axis
                  cutomize the scale of the axis
custom_ratio
                  the tick number of x_axis
x_ticks
y_ticks
                  the tick number of y_axis
z_ticks
                  the tick number of z_axis
show_plot
                  show plot
return_plot
                  return plot object
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### Value

plotly

### **Examples**

showPattern3D(gobject)

showPatternGenes

showPatternGenes

## Description

show genes correlated with spatial patterns

```
showPatternGenes(
  gobject,
  spatPatObj,
  dimension = 1,
  top_pos_genes = 5,
  top_neg_genes = 5,
  point_size = 1,
  return_DT = FALSE,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "showPatternGenes"
)
```

200 showProcessingSteps

## **Arguments**

gobject giotto object

spatPatObj Output from detectSpatialPatterns

dimension dimension to plot genes for.top\_pos\_genes Top positively correlated genes.top\_neg\_genes Top negatively correlated genes.

point\_size size of points

return\_DT if TRUE, it will return the data.table used to generate the plots

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function()

default\_save\_name

default save name for saving, don't change, change save\_name in save\_param

### Value

ggplot

### **Examples**

showPatternGenes(gobject)

showProcessingSteps

showProcessingSteps

## Description

shows the sequential processing steps that were performed in a summarized format

## Usage

showProcessingSteps(gobject)

## Arguments

gobject giotto object

#### Value

list of processing steps and names

## Examples

showProcessingSteps(gobject)

showSpatialCorGenes 201

showSpatialCorGenes showSpatialCorGenes

## Description

Shows and filters spatially correlated genes

## Usage

```
showSpatialCorGenes(
  spatCorObject,
  use_clus_name = NULL,
  selected_clusters = NULL,
  genes = NULL,
  min_spat_cor = 0.5,
  min_expr_cor = NULL,
  min_cor_diff = NULL,
  min_rank_diff = NULL,
  show_top_genes = NULL
)
```

### **Arguments**

```
spatCorObject
                  spatial correlation object
use_clus_name
                  cluster information to show
selected_clusters
                  subset of clusters to show
                  subset of genes to show
genes
                  filter on minimum spatial correlation
min_spat_cor
                  filter on minimum single-cell expression correlation
min_expr_cor
                  filter on minimum correlation difference (spatial vs expression)
min_cor_diff
                  filter on minimum correlation rank difference (spatial vs expression)
min_rank_diff
show_top_genes show top genes per gene
```

## Value

data.table with filtered information

### **Examples**

```
\verb|showSpatialCorGenes(gobject)|
```

202 showTopGeneToGene

showTopGeneToGene

show Top Gene To Gene

### **Description**

Show enriched/depleted gene-gene enrichments

### Usage

```
showTopGeneToGene(
  GTGscore,
  top_interactions = 10,
  direction = c("increased", "decreased"),
  complement_data = T,
  subset_cell_ints = NULL,
  subset_genes = NULL
)
```

## Arguments

### **Details**

Give more details ...

## Value

ggplot barplot

## **Examples**

```
showTopGeneToGene(scores)
```

signPCA 203

signPCA signPCA

## Description

identify significant prinicipal components (PCs)

### Usage

```
signPCA(
  gobject,
  method = c("screeplot", "jackstraw"),
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  genes_to_use = NULL,
  scale_unit = T,
  ncp = 50,
  scree_labels = T,
  scree_ylim = c(0, 10),
  jack_iter = 10,
  jack_threshold = 0.01,
  jack_verbose = T,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "signPCA",
)
```

### **Arguments**

```
gobject
                  giotto object
method
                  method to use to identify significant PCs
expression_values
                  expression values to use
                  cells or genes
reduction
genes_to_use
                  subset of genes to use for PCA
                  scale features before PCA
scale_unit
                  number of principal components to calculate
ncp
scree_labels
                  show labels on scree plot
                  y-axis limits on scree plot
scree_ylim
jack_iter
                  number of interations for jackstraw
jack_threshold p-value threshold to call a PC significant
                  show progress of jackstraw method
jack_verbose
show_plot
                  show plot
return_plot
                  return ggplot object
```

204 spatCellPlot

### **Details**

Two different methods can be used to assess the number of relevant or significant prinicipal components (PC's).

- 1. Screeplot works by plotting the explained variance of each individual PC in a barplot allowing you to identify which PC does not show a significant contribution anymore (= 'elbow method').
- 2. The Jackstraw method uses the permutationPA function. By systematically permuting genes it identifies robust, and thus significant, PCs.

multiple PCA results can be stored by changing the name parameter

#### Value

ggplot object for scree method and maxtrix of p-values for jackstraw

#### **Examples**

```
signPCA(gobject)
```

spatCellPlot

spatCellPlot

## Description

Visualize cells according to spatial coordinates

```
spatCellPlot(
  gobject,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
 cell_annotation_values,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
 gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
 point_size = 3,
 point_border_col = "black",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
  center_point_size = 4,
  center_point_border_col = "black",
```

spatCellPlot 205

```
center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  show_network = F,
  spatial_network_name = "spatial_network",
  network_color = NULL,
  network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
  grid_color = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 1,
  other_cells_alpha = 0.1,
  coord_fix_ratio = NULL,
  show_legend = T,
  legend_text = 8,
  axis_text = 8,
  axis_title = 8,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "spatCellPlot"
)
gobject
                giotto object
```

# **Arguments**

```
x-axis dimension name (default = 'sdimx')
sdimx
                  y-axis dimension name (default = 'sdimy')
sdimy
spat_enr_names names of spatial enrichment results to include
cell_annotation_values
                  numeric cell annotation columns
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
point_size
                  size of point (cell)
point_border_col
                  color of border around points
```

206 spatCellPlot

stroke size of border around points

```
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
show_network
                  show underlying spatial network
spatial_network_name
                  name of spatial network to use
                  color of spatial network
network_color
network_alpha
                  alpha of spatial network
                  show spatial grid
show_grid
spatial_grid_name
                  name of spatial grid to use
grid_color
                  color of spatial grid
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
show_legend
                  show legend
legend_text
                  size of legend text
                  size of axis text
axis_text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

point\_border\_stroke

### Value

ggplot

### **Examples**

```
spatCellPlot(gobject)
```

spatCellPlot2D

spatCellPlot2D

## **Description**

Visualize cells according to spatial coordinates

```
spatCellPlot2D(
 gobject,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
 cell_annotation_values,
 cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 select_cell_groups = NULL,
  select_cells = NULL,
 point_size = 3,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
 label_size = 4,
  label_fontface = "bold",
  show_network = F,
  spatial_network_name = "spatial_network",
 network_color = NULL,
 network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
  grid_color = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 other_cells_alpha = 0.1,
 coord_fix_ratio = NULL,
  show_legend = T,
  legend_text = 8,
  axis_text = 8,
 axis_title = 8,
  cow_n_col = 2,
```

```
cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatCellPlot2D"
Arguments
    gobject
                      giotto object
    sdimx
                      x-axis dimension name (default = 'sdimx')
    sdimy
                      y-axis dimension name (default = 'sdimy')
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                      numeric cell annotation columns
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
    select_cells
                      select subset of cells based on cell IDs
                      size of point (cell)
    point_size
    point_border_col
                      color of border around points
    point_border_stroke
                      stroke size of border around points
    show_cluster_center
                      plot center of selected clusters
    show_center_label
                      plot label of selected clusters
    center_point_size
                      size of center points
    label_size
                      size of labels
    label_fontface font of labels
    show_network
                      show underlying spatial network
    spatial_network_name
                      name of spatial network to use
                      color of spatial network
    network_color
    network_alpha
                      alpha of spatial network
    show_grid
                      show spatial grid
    spatial_grid_name
```

name of spatial grid to use

```
grid_color
                  color of spatial grid
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
show_legend
                  show legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

## **Details**

Description of parameters.

## Value

ggplot

### **Examples**

spatCellPlot2D(gobject)

spatDimCellPlot
 spatDimCellPlot

## Description

Visualize numerical features of cells according to spatial AND dimension reduction coordinates in 2D

```
spatDimCellPlot(
  gobject,
 plot_alignment = c("vertical", "horizontal"),
  spat_enr_names = NULL,
  cell_annotation_values,
  dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
  sdimx = "sdimx",
  sdimy = "sdimy",
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  dim_point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
  dim_show_center_label = T,
 dim_center_point_size = 4,
 dim_center_point_border_col = "black",
 dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_center_point_border_col = "black",
  spat_center_point_border_stroke = 0.1,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 nn_network_name = "sNN.pca",
  dim_edge_alpha = 0.5,
  spat_show_network = F,
  spatial_network_name = "spatial_network",
  spat_network_color = "red",
  spat_network_alpha = 0.5,
  spat_show_grid = F,
  spatial_grid_name = "spatial_grid",
  spat_grid_color = "green",
  show_other_cells = TRUE,
 other_cell_color = "grey";
  dim_other_point_size = 0.5,
  spat_other_point_size = 0.5,
```

```
spat_other_cells_alpha = 0.5,
      coord_fix_ratio = NULL,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_legend = T,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimCellPlot"
Arguments
   gobject
                     giotto object
    plot_alignment direction to align plot
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
   dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    sdimx
                     = spatial dimension to use on x-axis
    sdimy
                     = spatial dimension to use on y-axis
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    dim_point_size size of points in dim. reduction space
    dim_point_border_col
                     border color of points in dim. reduction space
    {\tt dim\_point\_border\_stroke}
                     border stroke of points in dim. reduction space
    spat_point_size
```

size of spatial points

spat\_point\_border\_col border color of spatial points spat\_point\_border\_stroke border stroke of spatial points dim\_show\_cluster\_center show the center of each cluster dim\_show\_center\_label provide a label for each cluster dim\_center\_point\_size size of the center point dim\_center\_point\_border\_col border color of center point dim\_center\_point\_border\_stroke stroke size of center point dim\_label\_size size of the center label dim\_label\_fontface font of the center label spat\_show\_cluster\_center show the center of each cluster spat\_show\_center\_label provide a label for each cluster spat\_center\_point\_size size of the center point spat\_label\_size size of the center label spat\_label\_fontface font of the center label show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) nn\_network\_name name of NN network to use, if show\_NN\_network = TRUE dim\_edge\_alpha column to use for alpha of the edges spat\_show\_network show spatial network spatial\_network\_name name of spatial network to use spat\_network\_color color of spatial network spat\_show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use spat\_grid\_color color of spatial grid show\_other\_cells display not selected cells

color of not selected cells

other\_cell\_color

```
dim_other_point_size
                  size of not selected dim cells
spat_other_point_size
                  size of not selected spat cells
spat_other_cells_alpha
                  alpha of not selected spat cells
coord_fix_ratio
                  ratio for coordinates
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
                  cowplot param: relative width
cow_rel_w
                  cowplot param: how to align
cow_align
show_legend
                  show legend
                  size of legend text
legend_text
axis_text
                  size of axis text
axis_title
                  size of axis title
                  show plot
show_plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

## **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

spatDimCellPlot(gobject)

spatDimCellPlot2D spatDimCellPlot2D

# Description

Visualize numerical features of cells according to spatial AND dimension reduction coordinates in 2D

```
spatDimCellPlot2D(
  gobject,
 plot_alignment = c("vertical", "horizontal"),
  spat_enr_names = NULL,
  cell_annotation_values,
  dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
  sdimx = "sdimx",
  sdimy = "sdimy",
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  dim_point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
  dim_show_center_label = T,
 dim_center_point_size = 4,
 dim_center_point_border_col = "black",
 dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_center_point_border_col = "black",
  spat_center_point_border_stroke = 0.1,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 nn_network_name = "sNN.pca",
  dim_edge_alpha = 0.5,
  spat_show_network = F,
  spatial_network_name = "spatial_network",
  spat_network_color = "red",
  spat_network_alpha = 0.5,
  spat_show_grid = F,
  spatial_grid_name = "spatial_grid",
  spat_grid_color = "green",
  show_other_cells = TRUE,
 other_cell_color = "grey";
  dim_other_point_size = 0.5,
  spat_other_point_size = 0.5,
```

```
spat_other_cells_alpha = 0.5,
      coord_fix_ratio = NULL,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_legend = T,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimCellPlot2D"
Arguments
   gobject
                     giotto object
    plot_alignment direction to align plot
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
   dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    sdimx
                     = spatial dimension to use on x-axis
    sdimy
                     = spatial dimension to use on y-axis
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    dim_point_size size of points in dim. reduction space
    dim_point_border_col
                     border color of points in dim. reduction space
    {\tt dim\_point\_border\_stroke}
                     border stroke of points in dim. reduction space
    spat_point_size
```

size of spatial points

spat\_point\_border\_col border color of spatial points spat\_point\_border\_stroke border stroke of spatial points dim\_show\_cluster\_center show the center of each cluster dim\_show\_center\_label provide a label for each cluster dim\_center\_point\_size size of the center point dim\_center\_point\_border\_col border color of center point dim\_center\_point\_border\_stroke stroke size of center point dim\_label\_size size of the center label dim\_label\_fontface font of the center label spat\_show\_cluster\_center show the center of each cluster spat\_show\_center\_label provide a label for each cluster spat\_center\_point\_size size of the center point spat\_label\_size size of the center label spat\_label\_fontface font of the center label show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) nn\_network\_name name of NN network to use, if show\_NN\_network = TRUE dim\_edge\_alpha column to use for alpha of the edges spat\_show\_network show spatial network spatial\_network\_name name of spatial network to use spat\_network\_color color of spatial network spat\_show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use spat\_grid\_color color of spatial grid show\_other\_cells display not selected cells

color of not selected cells

other\_cell\_color

spatDimGenePlot 217

```
dim_other_point_size
                  size of not selected dim cells
spat_other_point_size
                  size of not selected spat cells
spat_other_cells_alpha
                  alpha of not selected spat cells
coord_fix_ratio
                  ratio for coordinates
cow_n_col
                  cowplot param: how many columns
                  cowplot param: relative height
cow_rel_h
                  cowplot param: relative width
cow_rel_w
cow_align
                  cowplot param: how to align
                  show legend
show_legend
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

#### **Details**

Description of parameters.

## Value

ggplot

## **Examples**

spatDimCellPlot2D(gobject)

spatDimGenePlot
 spatDimGenePlot

### **Description**

Visualize cells according to spatial AND dimension reduction coordinates in ggplot mode

218 spatDimGenePlot

#### Usage

```
spatDimGenePlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  plot_alignment = c("vertical", "horizontal"),
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  show_NN_network = F,
  show_spatial_network = F,
  show_spatial_grid = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  edge_alpha_dim = NULL,
  scale_alpha_with_expression = FALSE,
  spatial_network_name = "spatial_network",
  spatial_grid_name = "spatial_grid",
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
  midpoint = 0,
  genes_high_color = "red",
  genes_mid_color = "white",
  genes_low_color = "blue",
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_legend = T,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "spatDimGenePlot"
)
gobject
                giotto object
expression_values
                gene expression values to use
```

#### **Arguments**

```
plot_alignment direction to align plot
genes
                 genes to show
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
```

spatDimGenePlot 219

```
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
point_size
                 size of point (cell)
dim_point_border_col
                 color of border around points
dim_point_border_stroke
                 stroke size of border around points
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
                 name of NN network to use, if show_NN_network = TRUE
network_name
edge_alpha_dim dim reduction plot: column to use for alpha of the edges
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
spatial_network_name
                 name of spatial network to use
spatial_grid_name
                 name of spatial grid to use
spat_point_size
                 spatial plot: point size
spat_point_border_col
                 color of border around points
spat_point_border_stroke
                 stroke size of border around points
                 size of point (cell)
midpoint
cow_n_col
                 cowplot param: how many columns
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
                 cowplot param: how to align
cow_align
                 show legend
show_legend
show_plot
                 show plots
return_plot
                 return ggplot object
save_plot
                 directly save the plot [boolean]
                 list of saving parameters from all_plots_save_function
save_param
default_save_name
                 default save name for saving, don't change, change save_name in save_param
dim_point_size dim reduction plot: point size
```

### **Details**

Description of parameters.

#### Value

ggplot

220 spatDimGenePlot2D

#### See Also

```
spatDimGenePlot3D
```

### **Examples**

```
spatDimGenePlot(gobject)
```

spatDimGenePlot2D

spatDimGenePlot2D

### **Description**

Visualize cells according to spatial AND dimension reduction coordinates in ggplot mode

```
spatDimGenePlot2D(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 plot_alignment = c("vertical", "horizontal"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
 point_size = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  show_NN_network = F,
  show\_spatial\_network = F,
  show_spatial_grid = F,
  nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 edge_alpha_dim = NULL,
  scale_alpha_with_expression = FALSE,
  spatial_network_name = "spatial_network",
  spatial_grid_name = "spatial_grid",
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 midpoint = 0,
 genes_high_color = "red",
 genes_mid_color = "white",
 genes_low_color = "blue",
 cow_n_col = 2,
 cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_legend = T,
  show_plot = NA,
  return_plot = NA,
```

spatDimGenePlot2D 221

```
save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimGenePlot2D"
    )
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
    plot_alignment direction to align plot
    genes
                     genes to show
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    point_size
                     size of point (cell)
    dim_point_border_col
                     color of border around points
    dim_point_border_stroke
                     stroke size of border around points
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    edge_alpha_dim dim reduction plot: column to use for alpha of the edges
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
    spatial_network_name
                     name of spatial network to use
    spatial_grid_name
                     name of spatial grid to use
    spat_point_size
                     spatial plot: point size
    spat_point_border_col
                     color of border around points
    spat_point_border_stroke
                     stroke size of border around points
    midpoint
                     size of point (cell)
    cow_n_col
                     cowplot param: how many columns
    cow_rel_h
                     cowplot param: relative height
    cow_rel_w
                     cowplot param: relative width
```

cowplot param: how to align

cow\_align

222 spatDimGenePlot3D

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

```
spatDimGenePlot3D
```

#### **Examples**

```
spatDimGenePlot2D(gobject)
```

```
spatDimGenePlot3D
```

# Description

Visualize cells according to spatial AND dimension reduction coordinates in ggplot mode

```
spatDimGenePlot3D(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 plot_alignment = c("horizontal", "vertical"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2\_to\_use = 2,
 dim3_to_use = NULL,
 sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  genes,
 cluster_column = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
```

spatDimGenePlot3D 223

```
other_cell_color = "lightgrey",
     other_point_size = 1.5,
      show_NN_network = F,
      nn_network_to_use = "sNN",
     network_name = "sNN.pca",
     label_size = 16,
      genes_low_color = "blue",
      genes_mid_color = "white",
     genes_high_color = "red",
     dim_point_size = 3,
     nn_network_alpha = 0.5,
      show_spatial_network = F,
      spatial_network_name = "spatial_network",
     network_color = "lightgray",
      spatial_network_alpha = 0.5,
      show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_grid_alpha = 0.5,
      spatial_point_size = 3,
     legend_text_size = 12,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
      x_{ticks} = NULL,
     y_ticks = NULL,
      z_ticks = NULL,
      show_plot = NA,
     return_plot = NA,
      save_plot = NA,
      save_param = list(),
     default_save_name = "spatDimGenePlot3D"
   )
Arguments
                    giotto object
   gobject
   expression_values
                    gene expression values to use
   plot_alignment direction to align plot
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
   dim1_to_use
                    dimension to use on x-axis
   dim2_to_use
                    dimension to use on y-axis
   dim3_to_use
                    dimension to use on z-axis
                    genes to show
   genes
   show_NN_network
```

show underlying NN network

show\_other\_cells = T,

```
nn_network_to_use ty
```

type of NN network to use (kNN vs sNN)

dim\_point\_size dim reduction plot: point size

spatial\_network\_name

name of spatial network to use

spatial\_grid\_name

name of spatial grid to use

spatial\_point\_size

spatial plot: point size

show\_plot show plots

return\_plot return plotly object

save\_plot directly save the plot [boolean]

save\_param list of saving parameters from all\_plots\_save\_function

 ${\tt default\_save\_name}$ 

default save name for saving, don't change, change save\_name in save\_param

edge\_alpha\_dim dim reduction plot: column to use for alpha of the edges

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

show\_legend show legend

#### **Details**

Description of parameters.

# Value

plotly

### **Examples**

spatDimGenePlot3D(gobject)

spatDimPlot

spatDimPlot

# Description

Visualize cells according to spatial AND dimension reduction coordinates 2D

```
spatDimPlot(
  gobject,
 plot_alignment = c("vertical", "horizontal"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
  dim1_to_use = 1,
 dim2\_to\_use = 2,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  dim_point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
  dim_show_cluster_center = F,
 dim_show_center_label = T,
 dim_center_point_size = 4,
  dim_center_point_border_col = "black",
  dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
  nn_network_alpha = 0.05,
  show_spatial_network = F,
  spat_network_name = "spatial_network",
  spat_network_color = "blue",
  spat_network_alpha = 0.5,
  show_spatial_grid = F,
  spat_grid_name = "spatial_grid",
  spat_grid_color = "blue",
  show_other_cells = T,
 other_cell_color = "lightgrey",
  dim_other_point_size = 1,
  spat_other_point_size = 1,
```

```
spat_other_cells_alpha = 0.5,
      dim_show_legend = F,
      spat_show_legend = F,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimPlot"
    )
Arguments
    gobject
                     giotto object
    plot_alignment direction to align plot
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    sdimx
                     = spatial dimension to use on x-axis
    sdimy
                     = spatial dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    dim_point_size size of points in dim. reduction space
    dim_point_border_col
                     border color of points in dim. reduction space
    dim_point_border_stroke
                     border stroke of points in dim. reduction space
    spat_point_size
                     size of spatial points
    spat_point_border_col
```

border color of spatial points

spat\_point\_border\_stroke border stroke of spatial points dim\_show\_cluster\_center show the center of each cluster dim\_show\_center\_label provide a label for each cluster dim\_center\_point\_size size of the center point dim\_center\_point\_border\_col border color of center point dim\_center\_point\_border\_stroke stroke size of center point dim\_label\_size size of the center label dim\_label\_fontface font of the center label spat\_show\_cluster\_center show the center of each cluster spat\_show\_center\_label provide a label for each cluster spat\_center\_point\_size size of the center point spat\_label\_size size of the center label spat\_label\_fontface font of the center label show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) name of NN network to use, if show\_NN\_network = TRUE network\_name nn\_network\_alpha column to use for alpha of the edges show\_spatial\_network show spatial network spat\_network\_name name of spatial network to use spat\_network\_color color of spatial network show\_spatial\_grid show spatial grid spat\_grid\_name name of spatial grid to use spat\_grid\_color color of spatial grid show\_other\_cells display not selected cells other\_cell\_color color of not selected cells

size of not selected dim cells

dim\_other\_point\_size

```
spat_other_point_size
                  size of not selected spat cells
spat_other_cells_alpha
                  alpha of not selected spat cells
dim_show_legend
                  show legend of dimension reduction plot
spat_show_legend
                  show legend of spatial plot
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

#### Value

ggplot

### See Also

spatDimPlot2D and spatDimPlot3D for 3D visualization.

# **Examples**

spatDimPlot(gobject)

spatDimPlot2D spatDimPlot2D

# Description

Visualize cells according to spatial AND dimension reduction coordinates 2D

```
spatDimPlot2D(
  gobject,
 plot_alignment = c("vertical", "horizontal"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
  dim1_to_use = 1,
 dim2\_to\_use = 2,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  dim_point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
  dim_show_cluster_center = F,
 dim_show_center_label = T,
 dim_center_point_size = 4,
  dim_center_point_border_col = "black",
  dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
  nn_network_alpha = 0.05,
  show_spatial_network = F,
  spat_network_name = "spatial_network",
  spat_network_color = "blue",
  spat_network_alpha = 0.5,
  show_spatial_grid = F,
  spat_grid_name = "spatial_grid",
  spat_grid_color = "blue",
  show_other_cells = T,
 other_cell_color = "lightgrey",
  dim_other_point_size = 1,
  spat_other_point_size = 1,
```

```
spat_other_cells_alpha = 0.5,
      dim_show_legend = F,
      spat_show_legend = F,
      legend_text = 8,
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimPlot2D"
    )
Arguments
    gobject
                      giotto object
    plot_alignment direction to align plot
    dim_reduction_to_use
                      dimension reduction to use
    dim_reduction_name
                      dimension reduction name
    dim1_to_use
                      dimension to use on x-axis
    dim2_to_use
                      dimension to use on y-axis
    sdimx
                      = spatial dimension to use on x-axis
    sdimy
                      = spatial dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_color
                      color for cells (see details)
    color_as_factor
                      convert color column to factor
    cell_color_code
                      named vector with colors
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
    select_cells
                      select subset of cells based on cell IDs
    dim_point_size size of points in dim. reduction space
    dim_point_border_col
                      border color of points in dim. reduction space
    dim_point_border_stroke
                      border stroke of points in dim. reduction space
    spat_point_size
                      size of spatial points
    spat_point_border_col
                      border color of spatial points
```

spat\_point\_border\_stroke border stroke of spatial points dim\_show\_cluster\_center show the center of each cluster dim\_show\_center\_label provide a label for each cluster dim\_center\_point\_size size of the center point dim\_center\_point\_border\_col border color of center point dim\_center\_point\_border\_stroke stroke size of center point dim\_label\_size size of the center label dim\_label\_fontface font of the center label spat\_show\_cluster\_center show the center of each cluster spat\_show\_center\_label provide a label for each cluster spat\_center\_point\_size size of the center point spat\_label\_size size of the center label spat\_label\_fontface font of the center label show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) name of NN network to use, if show\_NN\_network = TRUE network\_name nn\_network\_alpha column to use for alpha of the edges show\_spatial\_network show spatial network spat\_network\_name name of spatial network to use spat\_network\_color color of spatial network show\_spatial\_grid show spatial grid spat\_grid\_name name of spatial grid to use spat\_grid\_color color of spatial grid show\_other\_cells display not selected cells other\_cell\_color color of not selected cells dim\_other\_point\_size

size of not selected dim cells

```
spat_other_point_size
                  size of not selected spat cells
spat_other_cells_alpha
                  alpha of not selected spat cells
dim_show_legend
                  show legend of dimension reduction plot
spat_show_legend
                  show legend of spatial plot
legend_text
                  size of legend text
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

#### Value

ggplot

### See Also

spatDimPlot3D

# **Examples**

spatDimPlot2D(gobject)

spatDimPlot3D spatDimPlot3D

# Description

Visualize cells according to spatial AND dimension reduction coordinates in plotly mode

```
spatDimPlot3D(
 gobject,
 plot_alignment = c("horizontal", "vertical"),
  dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2\_to\_use = 2,
 dim3_to_use = 3,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = 16,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1.5,
 cell_color = NULL,
 color_as_factor = T,
  cell_color_code = NULL,
  dim_point_size = 3,
 nn_network_alpha = 0.5,
  show\_spatial\_network = F,
  spatial_network_name = "spatial_network",
  network_color = "lightgray",
  spatial_network_alpha = 0.5,
  show_spatial_grid = F,
  spatial_grid_name = "spatial_grid",
  spatial_grid_color = NULL,
  spatial_grid_alpha = 0.5,
  spatial_point_size = 3,
  axis_scale = c("cube", "real", "custom"),
 custom_ratio = NULL,
 x_ticks = NULL,
 y_ticks = NULL,
 z_ticks = NULL,
 legend_text_size = 12,
  show_plot = NA,
 return_plot = NA,
  save_plot = NA,
  save_param = list(),
 default_save_name = "spatDimPlot3D"
```

### Arguments

gobject giotto object plot\_alignment direction to align plot dim\_reduction\_to\_use dimension reduction to use dim\_reduction\_name dimension reduction name dimension to use on x-axis dim1\_to\_use dim2\_to\_use dimension to use on y-axis dim3\_to\_use dimension to use on z-axis sdimx = spatial dimension to use on x-axis = spatial dimension to use on y-axis sdimy sdimz = spatial dimension to use on z-axis show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) name of NN network to use, if show\_NN\_network = TRUE network\_name show\_cluster\_center show the center of each cluster show\_center\_label provide a label for each cluster center\_point\_size size of the center point size of the center label label\_size select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells cell\_color color for cells (see details) color\_as\_factor convert color column to factor cell\_color\_code named vector with colors dim\_point\_size size of points in dim. reduction space nn\_network\_alpha column to use for alpha of the edges show\_spatial\_network show spatial network spatial\_network\_name name of spatial network to use

```
spatial_network_alpha
                  alpha of spatial network
show_spatial_grid
                  show spatial grid
spatial_grid_name
                  name of spatial grid to use
spatial_grid_color
                  color of spatial grid
spatial_point_size
                  size of spatial points
show_plot
                 show plot
return_plot
                 return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
dim_point_border_col
                  border color of points in dim. reduction space
dim_point_border_stroke
                  border stroke of points in dim. reduction space
spatial_network_color
                  color of spatial network
spatial_other_point_size
                  size of not selected spatial points
spatial_other_cells_alpha
                  alpha of not selected spatial points
dim_other_point_size
                  size of not selected dim. reduction points
```

#### **Details**

Description of parameters.

show legend

show\_legend

## Value

plotly

# **Examples**

spatDimPlot3D(gobject)

236 spatGenePlot

spatGenePlot

spatGenePlot

### **Description**

Visualize cells and gene expression according to spatial coordinates

# Usage

```
spatGenePlot(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 genes,
 genes_high_color = "darkred",
  genes_mid_color = "white",
 genes_low_color = "darkblue",
 show_network = F,
 network_color = NULL,
 spatial_network_name = "spatial_network",
 edge_alpha = NULL,
  show\_grid = F,
 grid_color = NULL,
  spatial_grid_name = "spatial_grid",
 midpoint = 0,
  scale_alpha_with_expression = FALSE,
 point_size = 1,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_legend = T,
  cow_n_col = 2,
 cow_rel_h = 1,
 cow_rel_w = 1,
 cow_align = "h",
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatGenePlot"
)
```

### **Arguments**

```
gobject giotto object
expression_values
gene expression values to use
genes genes to show
genes_high_color
color represents high gene expression
genes_mid_color
color represents middle gene expression
```

spatGenePlot 237

```
genes_low_color
                  color represents low gene expression
show_network
                  show underlying spatial network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
midpoint
                  expression midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  size of point (cell)
point_size
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plots
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  parameters for cowplot::save_plot()
. . .
```

#### **Details**

Description of parameters.

#### Value

ggplot

### See Also

 ${\tt spatGenePlot3D} \ and \ {\tt spatGenePlot2D}$ 

## **Examples**

```
spatGenePlot(gobject)
```

238 spatGenePlot2D

spatGenePlot2D

spatGenePlot2D

### **Description**

Visualize cells and gene expression according to spatial coordinates

### Usage

```
spatGenePlot2D(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 genes,
 genes_high_color = "darkred",
  genes_mid_color = "white",
 genes_low_color = "darkblue",
 show_network = F,
 network_color = NULL,
 spatial_network_name = "spatial_network",
 edge_alpha = NULL,
  show\_grid = F,
 grid_color = NULL,
  spatial_grid_name = "spatial_grid",
 midpoint = 0,
  scale_alpha_with_expression = FALSE,
 point_size = 1,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_legend = T,
  cow_n_col = 2,
 cow_rel_h = 1,
 cow_rel_w = 1,
 cow_align = "h",
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatGenePlot2D"
)
```

### **Arguments**

```
gobject giotto object
expression_values
gene expression values to use
genes genes to show
genes_high_color
color represents high gene expression
genes_mid_color
color represents middle gene expression
```

spatGenePlot2D 239

```
genes_low_color
                  color represents low gene expression
show_network
                  show underlying spatial network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
midpoint
                  expression midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  size of point (cell)
point_size
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plots
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  parameters for cowplot::save_plot()
. . .
```

#### **Details**

Description of parameters.

#### Value

ggplot

### See Also

 ${\tt spatGenePlot3D}$ 

## **Examples**

spatGenePlot2D(gobject)

240 spatGenePlot3D

spatGenePlot3D spatGenePlot3D

### **Description**

Visualize cells and gene expression according to spatial coordinates

#### Usage

```
spatGenePlot3D(
 gobject,
  expression_values = c("normalized", "scaled", "custom"),
 genes,
 show_network = F,
 network_color = NULL,
  spatial_network_name = "spatial_network",
 edge_alpha = NULL,
 show\_grid = F,
 cluster_column = NULL,
  select_cell_groups = NULL,
 select_cells = NULL,
 show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 genes_high_color = NULL,
 genes_mid_color = "white",
 genes_low_color = "blue",
  spatial_grid_name = "spatial_grid",
 point_size = 2,
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
 x_ticks = NULL,
 y_ticks = NULL,
 z_ticks = NULL,
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatGenePlot3D"
)
```

# **Arguments**

```
gobject giotto object
expression_values
gene expression values to use
genes genes to show
show_network show underlying spatial network
network_color color of spatial network
```

Spatial\_AEH 241

```
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
genes_high_color
                  color represents high gene expression
genes_mid_color
                  color represents middle gene expression
genes_low_color
                  color represents low gene expression
spatial_grid_name
                  name of spatial grid to use
                  size of point (cell)
point_size
{\sf show\_legend}
                  show legend
show_plot
                  show plots
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
grid_color
                  color of spatial grid
midpoint
                  expression midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  parameters for cowplot::save_plot()
. . .
```

# Details

Description of parameters.

#### Value

ggplot

# Examples

spatGenePlot3D(gobject)

Spatial\_AEH Spatial\_AEH

# Description

calculate automatic expression histology with spatialDE method

242 Spatial\_DE

### Usage

```
Spatial_AEH(
  gobject = NULL,
  results = NULL,
  pattern_num = 5,
  1 = 1.05,
  show\_AEH = T,
  sdimx = NULL,
  sdimy = NULL,
  point_size = 3,
  point_alpha = 1,
  low_color = "blue",
  mid_color = "white",
  high_color = "red",
  midpoint = 0,
  python_path = NULL
)
```

# Arguments

gobject Giotto object
results output from spatial\_DE
pattern\_num the number of gene expression patterns

show\_AEH show AEH plot

python\_path specify specific path to python if required

### **Details**

Description.

#### Value

a list or a dataframe of SVs

### **Examples**

```
Spatial_AEH(gobject)
```

Spatial\_DE Spatial\_DE

# Description

calculate spatial varible genes with spatialDE method

spatPlot 243

### Usage

```
Spatial_DE(
  gobject = NULL,
  show_plot = T,
  size = c(4, 2, 1),
  color = c("blue", "green", "red"),
  sig_alpha = 0.5,
  unsig_alpha = 0.5,
  python_path = NULL
)
```

#### **Arguments**

gobject Giotto object
show\_plot show FSV plot
python\_path specific path to python if required

### **Details**

Description.

#### Value

a list or a dataframe of SVs

### **Examples**

Spatial\_DE(gobject)

spatPlot

spatPlot

# Description

Visualize cells according to spatial coordinates

```
spatPlot(
  gobject,
  group_by = NULL,
  group_by_subset = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
```

244 spatPlot

```
select_cell_groups = NULL,
select_cells = NULL,
point_size = 3,
point_border_col = "black",
point_border_stroke = 0.1,
show_cluster_center = F,
show_center_label = F,
center_point_size = 4,
center_point_border_col = "black",
center_point_border_stroke = 0.1,
label_size = 4,
label_fontface = "bold",
show_network = F,
spatial_network_name = "spatial_network",
network_color = NULL,
network\_alpha = 1,
show_grid = F,
spatial_grid_name = "spatial_grid",
grid_color = NULL,
show_other_cells = T,
other_cell_color = "lightgrey",
other_point_size = 1,
other_cells_alpha = 0.1,
coord_fix_ratio = NULL,
title = NULL,
show_legend = T,
legend_text = 8,
axis_text = 8,
axis_title = 8,
cow_n_col = 2,
cow_rel_h = 1,
cow_rel_w = 1,
cow_align = "h",
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spatPlot"
```

### **Arguments**

spatPlot 245

cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select subset of cells based on cell IDs select\_cells point\_size size of point (cell) point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network network\_alpha alpha of spatial network show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use grid\_color color of spatial grid show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size point size of not selected cells other\_cells\_alpha alpha of not selected cells coord\_fix\_ratio fix ratio between x and y-axis title title of plot show\_legend show legend

legend\_text

size of legend text

246 spatPlot2D

```
size of axis text
axis_text
                  size of axis title
axis_title
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
groub_by
                  create multiple plots based on cell annotation column
```

#### **Details**

Description of parameters.

### Value

ggplot

#### See Also

```
spatPlot3D
```

### **Examples**

```
spatPlot(gobject)
```

spatPlot2D

spatPlot2D

# Description

Visualize cells according to spatial coordinates

```
spatPlot2D(
  gobject,
  group_by = NULL,
  group_by_subset = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
```

spatPlot2D 247

```
cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
 gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
 point_size = 3,
 point_border_col = "black",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
  center_point_size = 4,
 center_point_border_col = "black",
  center_point_border_stroke = 0.1,
 label_size = 4,
  label_fontface = "bold",
  show_network = F,
  spatial_network_name = "spatial_network",
 network_color = NULL,
 network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
 grid_color = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 other_cells_alpha = 0.1,
  coord_fix_ratio = NULL,
  title = NULL,
  show_legend = T,
 legend_text = 8,
 axis_text = 8,
 axis_title = 8,
 cow_n_col = 2,
 cow_rel_h = 1,
 cow_rel_w = 1,
 cow_align = "h",
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatPlot2D"
)
```

#### **Arguments**

248 spatPlot2D

color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select subset of cells based on cell IDs select\_cells point\_size size of point (cell) point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network network\_alpha alpha of spatial network show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use color of spatial grid grid\_color show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size point size of not selected cells other\_cells\_alpha alpha of not selected cells coord\_fix\_ratio fix ratio between x and y-axis

title of plot

title

spatPlot2D\_single 249

```
show_legend
                  show legend
legend_text
                  size of legend text
axis\_text
                  size of axis text
axis_title
                  size of axis title
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
groub_by
                  create multiple plots based on cell annotation column
```

#### **Details**

Description of parameters.

### Value

ggplot

### See Also

spatPlot3D

### **Examples**

spatPlot2D(gobject)

### **Description**

Visualize cells according to spatial coordinates

250 spatPlot2D\_single

#### Usage

```
spatPlot2D_single(
 gobject,
 sdimx = "sdimx",
 sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
 color_as_factor = T,
 cell_color_code = NULL,
 cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 select_cell_groups = NULL,
  select_cells = NULL,
 point_size = 3,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
 center_point_size = 4,
 center_point_border_col = "black",
  center_point_border_stroke = 0.1,
 label_size = 4,
 label_fontface = "bold",
  show_network = F,
  spatial_network_name = "spatial_network",
 network_color = NULL,
 network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
 grid_color = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 other_cells_alpha = 0.1,
  coord_fix_ratio = NULL,
  title = NULL,
  show_legend = T,
 legend_text = 8,
 axis_text = 8,
 axis_title = 8,
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatPlot2D_single"
)
```

### **Arguments**

```
gobject giotto object

sdimx x-axis dimension name (default = 'sdimx')
```

spatPlot2D\_single 251

sdimy y-axis dimension name (default = 'sdimy') spat\_enr\_names names of spatial enrichment results to include color for cells (see details) cell\_color color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell color parameter select subset of cells based on cell IDs select\_cells size of point (cell) point\_size point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network network\_alpha alpha of spatial network show\_grid show spatial grid  ${\tt spatial\_grid\_name}$ name of spatial grid to use color of spatial grid grid\_color show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size point size of not selected cells other\_cells\_alpha alpha of not selected cells

252 spatPlot3D

```
coord_fix_ratio
                  fix ratio between x and y-axis
                  title of plot
title
                  show legend
show_legend
                  size of legend text
legend_text
axis_text
                  size of axis text
                  size of axis title
axis_title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

#### Value

ggplot

### See Also

spatPlot3D

### **Examples**

```
spatPlot2D_single(gobject)
```

spatPlot3D spatPlot3D

# Description

Visualize cells according to spatial coordinates

```
spatPlot3D(
  gobject,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  point_size = 3,
  cell_color = NULL,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
```

spatPlot3D 253

```
show_other_cells = T,
other_cell_color = "lightgrey",
other_point_size = 0.5,
show_network = F,
network_color = NULL,
network_alpha = 1,
other_cell_alpha = 0.5,
spatial_network_name = "spatial_network",
show\_grid = F,
grid_color = NULL,
spatial_grid_name = "spatial_grid",
title = "",
show_legend = T,
axis_scale = c("cube", "real", "custom"),
custom_ratio = NULL,
x_ticks = NULL,
y_ticks = NULL,
z_{ticks} = NULL,
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spat3D"
```

```
gobject
                  giotto object
sdimx
                  x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
sdimz
                  z-axis dimension name (default = 'sdimy')
point_size
                  size of point (cell)
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
show_network
                  show underlying spatial network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
```

```
spatial_grid_name
                  name of spatial grid to use
title
                  title of plot
show_legend
                  show legend
axis_scale
                  the way to scale the axis
custom_ratio
                  customize the scale of the plot
                  set the number of ticks on the x-axis
x_ticks
                  set the number of ticks on the y-axis
y_ticks
z_ticks
                  set the number of ticks on the z-axis
                  show plot
show_plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

### Value

ggplot

# **Examples**

```
spatPlot3D(gobject)
```

```
specific Cell Cell communication Scores\\ specific Cell Cell communication Scores
```

# **Description**

Specific Cell-Cell communication scores based on spatial expression of interacting cells

```
specificCellCellcommunicationScores(
  gobject,
  spatial_network_name = "spatial_network",
  cluster_column = "cell_types",
  random_iter = 100,
  cell_type_1 = "astrocyte",
  cell_type_2 = "endothelial",
  gene_set_1,
  gene_set_2,
```

```
log2FC_addendum = 0.1,
min_observations = 2,
verbose = T
)
```

# **Arguments**

```
gobject
                  giotto object to use
spatial_network_name
                  spatial network to use for identifying interacting cells
cluster_column cluster column with cell type information
random_iter
                  number of iterations
                  first cell type
cell_type_1
cell_type_2
                  second cell type
gene_set_1
                  first specific gene set from gene pairs
                  second specific gene set from gene pairs
gene_set_2
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
verbose
                  verbose
```

### **Details**

Statistical framework to identify if pairs of genes (such as ligand-receptor combinations) are expressed at higher levels than expected based on a reshuffled null distribution of gene expression values in cells that are spatially in proximity to eachother. More details will follow soon.

# Value

Cell-Cell communication scores for gene pairs based on spatial interaction

#### **Examples**

```
{\tt specific Cell Cell communication Scores (gobject)}
```

# Description

Merge selected clusters based on pairwise correlation scores and size of cluster.

```
split_dendrogram_in_two(dend)
```

256 stitchFieldCoordinates

### **Arguments**

dend dendrogram object

### Value

list of two dendrograms and height of node

### **Examples**

```
split_dendrogram_in_two(dend)
```

stitchFieldCoordinates

stitchFieldCoordinates

# **Description**

Helper function to stitch field coordinates together to form one complete picture

# Usage

```
stitchFieldCoordinates(
  location_file,
  offset_file,
  cumulate_offset_x = F,
  cumulate_offset_y = F,
  field_col = "Field of View",
  X_coord_col = "X",
  Y_coord_col = "Y",
  reverse_final_x = F,
  reverse_final_y = T
)
```

```
location_file location dataframe with X and Y coordinates
offset_file
                  dataframe that describes the offset for each field (see details)
cumulate_offset_x
                  (boolean) Do the x-axis offset values need to be cumulated?
cumulate_offset_y
                  (boolean) Do the y-axis offset values need to be cumulated?
field_col
                  column that indicates the field within the location_file
X_coord_col
                  column that indicates the x coordinates
Y_coord_col
                  column that indicates the x coordinates
reverse_final_x
                  (boolean) Do the final x coordinates need to be reversed?
reverse_final_y
                  (boolean) Do the final y coordinates need to be reversed?
```

subClusterCells 257

#### **Details**

Stitching of fields:

- 1. have cell locations: at least 3 columns: field, X, Y
- 2. create offset file: offset file has 3 columns: field, x\_offset, y\_offset
- 3. create new cell location file by stitching original cell locations with stitchFieldCoordinates
- 4. provide new cell location file to createGiottoObject

### Value

Updated location dataframe with new X ['X\_final'] and Y ['Y\_final'] coordinates

### **Examples**

```
stitchFieldCoordinates(gobject)
```

subClusterCells

subClusterCells

# **Description**

subcluster cells

```
subClusterCells(
 gobject,
 name = "sub_clus",
 cluster_method = c("leiden", "louvain_community", "louvain_multinet"),
  cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
    = "normalized"),
 hvg_min_perc_cells = 5,
 hvg_mean_expr_det = 1,
 use_all_genes_as_hvg = FALSE,
 min_nr_of_hvg = 5,
 pca_param = list(expression_values = "normalized", scale_unit = T),
 nn_param = list(dimensions_to_use = 1:20),
 k_neighbors = 10,
 resolution = 1,
 gamma = 1,
 omega = 1,
 python_path = NULL,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 return_gobject = TRUE,
  verbose = T
)
```

258 subClusterCells

#### **Arguments**

gobject giotto object

name name for new clustering result cluster\_method clustering method to use cluster\_column to subcluster

selected\_clusters

only do subclustering on these clusters

hvg\_param parameters for calculateHVG

hvg\_min\_perc\_cells

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

forces all genes to be HVG and to be used as input for PCA

min\_nr\_of\_hvg minimum number of HVG, or all genes will be used as input for PCA

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

resolution resolution gamma gamma omega omega

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

return\_gobject boolean: return giotto object (default = TRUE)

verbose verbose

#### **Details**

This function performs subclustering on selected clusters. The systematic steps are:

- 1. subset Giotto object
- 2. identify highly variable genes
- 3. run PCA
- 4. create nearest neighbouring network
- 5. do clustering

### Value

giotto object with new subclusters appended to cell metadata

### See Also

doLouvainCluster\_multinet, doLouvainCluster\_community and @seealso doLeidenCluster

### **Examples**

subClusterCells(gobject)

subsetGiotto 259

subsetGiotto

subsetGiot to

# Description

subsets Giotto object including previous analyses.

# Usage

```
subsetGiotto(gobject, cell_ids = NULL, gene_ids = NULL, verbose = FALSE)
```

# Arguments

```
gobject giotto object
cell_ids cell IDs to keep
gene_ids gene IDs to keep
verbose be verbose
```

### Value

giotto object

# **Examples**

```
subsetGiotto(gobject)
```

 ${\tt subsetGiottoLocs}$ 

subsetGiottoLocs

# **Description**

subsets Giotto object based on spatial locations

```
subsetGiottoLocs(
  gobject,
  x_max = NULL,
  x_min = NULL,
  y_max = NULL,
  y_min = NULL,
  z_max = NULL,
  z_min = NULL,
  return_gobject = T,
  verbose = FALSE
)
```

260 viewHMRFresults

# **Arguments**

gobject	giotto object
x_max	maximum x-coordinate
x_min	minimum x-coordinate
y_max	maximum y-coordinate
y_min	minimum y-coordinate
z_max	maximum z-coordinate
z_min	minimum z-coordinate
return_gobject	return Giotto object

# **Details**

if return\_gobject = FALSE, then a filtered combined metadata data.table will be returned

# Value

giotto object

# **Examples**

```
subsetGiottoLocs(gobject)
```

viewHMRFresults

viewHMRFresults

# Description

View results from doHMRF.

# Usage

```
viewHMRFresults(
  gobject,
  HMRFoutput,
  k = NULL,
  betas_to_view = NULL,
  third_dim = NULL,
  ...
)
```

```
gobject giotto object

HMRFoutput HMRF output from doHMRF

k number of HMRF domains

betas_to_view results from different betas that you want to view

... paramters to visPlot()
```

viewHMRFresults2D 261

### **Details**

Description ...

# Value

spatial plots with HMRF domains

# See Also

visPlot

### **Examples**

viewHMRFresults(gobject)

viewHMRFresults2D

viewHMRFresults2D

# Description

View results from doHMRF.

# Usage

```
viewHMRFresults2D(
  gobject,
  HMRFoutput,
  k = NULL,
  betas_to_view = NULL,
  third_dim = NULL,
  ...
)
```

# Arguments

gobject giotto object

HMRF output from doHMRF k number of HMRF domains

 $\verb|betas_to_view| \quad \textit{results from different betas that you want to view}$ 

... paramters to visPlot()

# **Details**

Description ...

### Value

spatial plots with HMRF domains

### See Also

```
spatPlot2D
```

262 viewHMRFresults3D

# **Examples**

```
viewHMRFresults2D(gobject)
```

viewHMRFresults3D

viewHMRFresults3D

# Description

View results from doHMRF.

# Usage

```
viewHMRFresults3D(
  gobject,
  HMRFoutput,
  k = NULL,
  betas_to_view = NULL,
  third_dim = NULL,
  ...
)
```

# Arguments

```
gobject giotto object
```

HMRF output from doHMRF k number of HMRF domains

betas\_to\_view results from different betas that you want to view

... paramters to visPlot()

# **Details**

Description ...

# Value

spatial plots with HMRF domains

### See Also

```
spatPlot3D
```

# **Examples**

```
viewHMRFresults3D(gobject)
```

violinPlot 263

violinPlot

violinPlot

# **Description**

Creates violinplot for selected clusters

# Usage

```
violinPlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  cluster_column,
  cluster_custom_order = NULL,
  color_violin = c("genes", "cluster"),
  cluster_color_code = NULL,
  strip_position = c("top", "right", "left", "bottom"),
  strip\_text = 7,
  axis_text_x_size = 10,
  axis_text_y_size = 6,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "violinPlot"
)
```

```
gobject
                  giotto object
expression_values
                  expression values to use
                  genes to plot
genes
cluster_column name of column to use for clusters
cluster_custom_order
                  custom order of clusters
color_violin
                  color violin according to genes or clusters
cluster_color_code
                  color code for clusters
strip_position position of gene labels
strip_text
                  size of strip text
\verb"axis_text_x_size"
                  size of x-axis text
axis_text_y_size
                  size of y-axis text
show_plot
                  show plot
return_plot
                  return ggplot object
```

264 visDimGenePlot

### Value

ggplot

### **Examples**

```
violinPlot(gobject)
```

visDimGenePlot

visDimGenePlot

# **Description**

Visualize cells and gene expression according to dimension reduction coordinates

```
visDimGenePlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  dim3_to_use = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  scale_alpha_with_expression = FALSE,
  point_size = 1,
  genes_high_color = NULL,
  genes_mid_color = "white",
  genes_low_color = "blue",
  point_border_col = "black",
  point_border_stroke = 0.1,
  midpoint = 0,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h"
  show_legend = T,
  plot_method = c("ggplot", "plotly"),
  show_plots = F
)
```

visDimGenePlot 265

### **Arguments**

gobject giotto object

expression\_values

gene expression values to use

genes genes to show

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimension reduction name

dim1\_to\_use dimension to use on x-axis dim2\_to\_use dimension to use on y-axis dim3\_to\_use dimension to use on z-axis

show\_NN\_network

show underlying NN network

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

 $network\_name$  name of NN network to use, if  $show\_NN\_network = TRUE$ 

edge\_alpha column to use for alpha of the edges

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

midpoint size of point (cell)

cow\_n\_col cowplot param: how many columns
cow\_rel\_h cowplot param: relative height
cow\_rel\_w cowplot param: relative width
cow\_align cowplot param: how to align

show\_legend show legend show\_plots show plots

# Details

Description of parameters.

#### Value

ggplot

# Examples

visDimGenePlot(gobject)

# **Description**

Visualize cells and gene expression according to dimension reduction coordinates

# Usage

```
visDimGenePlot_2D_ggplot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  scale_alpha_with_expression = FALSE,
  point_size = 1,
  genes_high_color = "red",
  genes_mid_color = "white",
  genes_low_color = "blue",
  point_border_col = "black",
  point_border_stroke = 0.1,
  midpoint = 0,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_legend = T,
  show_plots = F
```

```
gobject giotto object
expression_values
gene expression values to use
genes genes to show
dim_reduction_to_use
dimension reduction to use
dim_reduction_name
dimension reduction name
dim1_to_use
dimension to use on x-axis
```

```
dim2_to_use
                 dimension to use on y-axis
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
                 name of NN network to use, if show_NN_network = TRUE
network_name
edge_alpha
                 column to use for alpha of the edges
scale_alpha_with_expression
                 scale expression with ggplot alpha parameter
                 size of point (cell)
point_size
point_border_col
                 color of border around points
point_border_stroke
                 stroke size of border around points
                 size of point (cell)
midpoint
cow_n_col
                 cowplot param: how many columns
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
cow_align
                 cowplot param: how to align
show_legend
                 show legend
show_plots
                 show plots
```

### **Details**

Description of parameters.

# Value

ggplot

# **Examples**

```
visDimGenePlot_2D_ggplot(gobject)
```

# Description

Visualize cells and gene expression according to dimension reduction coordinates

#### **Usage**

```
visDimGenePlot_3D_plotly(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  dim3_to_use = 3,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  point_size = 1,
  genes_high_color = NULL,
  genes_mid_color = "white",
  genes_low_color = "blue",
  show_legend = T,
  show_plots = F
)
```

# **Arguments**

```
gobject
                 giotto object
expression_values
                 gene expression values to use
genes
                 genes to show
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
dim3_to_use
                 dimension to use on z-axis
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
network_name
                 name of NN network to use, if show_NN_network = TRUE
edge_alpha
                 column to use for alpha of the edges
point_size
                 size of point (cell)
show_legend
                 show legend
                 show plots
show_plots
```

### **Details**

Description of parameters.

visDimPlot 269

### Value

ggplot

#### **Examples**

visDimGenePlot\_3D\_plotly(gobject)

visDimPlot

visDimPlot

### **Description**

Visualize cells according to dimension reduction coordinates

```
visDimPlot(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  dim3_to_use = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  point_size = 3,
  point_border_col = "black",
  point_border_stroke = 0.1,
  plot_method = c("ggplot", "plotly"),
  show_legend = T,
  show_plot = F,
  return_plot = TRUE,
  save_plot = F,
  save_dir = NULL,
```

visDimPlot

```
save_folder = NULL,
      save_name = NULL,
      save_format = NULL,
      show_saved_plot = F,
    )
Arguments
                     giotto object
    gobject
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
                     dimension to use on z-axis
    dim3_to_use
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    show_cluster_center
                     plot center of selected clusters
    show_center_label
                     plot label of selected clusters
    center_point_size
                     size of center points
    label_size
                     size of labels
    label_fontface font of labels
    edge_alpha
                     column to use for alpha of the edges
    point_size
                     size of point (cell)
    point_border_col
                     color of border around points
    point_border_stroke
                     stroke size of border around points
    show_legend
                     show legend
    show_plot
                     show plot
    return_plot
                     return ggplot object
                     directly save the plot [boolean]
    save_plot
```

directory to save the plot

save\_dir

visDimPlot\_2D\_ggplot 271

### **Details**

Description of parameters.

### Value

ggplot or plotly

# **Examples**

```
visDimPlot(gobject)
```

```
visDimPlot_2D_ggplot visDimPlot_2D_ggplot
```

# Description

Visualize cells according to dimension reduction coordinates

```
visDimPlot_2D_ggplot(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
```

```
edge_alpha = NULL,
      point_size = 1,
      point_border_col = "black",
      point_border_stroke = 0.1,
      show_legend = T,
      show_plot = F,
      return_plot = TRUE,
      save_plot = F,
      save_dir = NULL,
      save_folder = NULL,
      save_name = NULL,
      save_format = NULL,
      show_saved_plot = F,
    )
Arguments
    gobject
                     giotto object
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    show_other_cells
                     display not selected cells
    other_cell_color
                     color of not selected cells
    other_point_size
                     size of not selected cells
    show_cluster_center
                     plot center of selected clusters
    show_center_label
                     plot label of selected clusters
    center_point_size
                     size of center points
```

visDimPlot\_2D\_plotly 273

#### **Details**

Description of parameters.

#### Value

ggplot

# **Examples**

```
visDimPlot_2D_ggplot(gobject)
```

```
visDimPlot_2D_plotly visDimPlot_2D_plotly
```

# **Description**

Visualize cells according to dimension reduction coordinates

```
visDimPlot_2D_plotly(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  color_as_factor = T,
  cell_color = NULL,
  cell_color_code = NULL,
  show_cluster_center = F,
  show_center_label = T,
```

```
center_point_size = 4,
label_size = 4,
edge_alpha = NULL,
point_size = 5
```

# **Arguments**

```
gobject
                 giotto object
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
                 dimension to use on x-axis
dim1_to_use
dim2_to_use
                 dimension to use on y-axis
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
network_name
                 name of NN network to use, if show_NN_network = TRUE
color_as_factor
                 convert color column to factor
cell_color
                 color for cells (see details)
cell_color_code
                 named vector with colors
show_cluster_center
                 plot center of selected clusters
show_center_label
                 plot label of selected clusters
center_point_size
                 size of center points
label_size
                 size of labels
edge_alpha
                 column to use for alpha of the edges
point_size
                 size of point (cell)
```

### **Details**

Description of parameters.

### Value

plotly

# **Examples**

```
visDimPlot_2D_plotly(gobject)
```

```
visDimPlot_3D_plotly
```

# **Description**

Visualize cells according to dimension reduction coordinates

# Usage

```
visDimPlot_3D_plotly(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  dim3_to_use = 3,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  color_as_factor = T,
  cell_color = NULL,
  cell_color_code = NULL,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = 4,
  edge_alpha = NULL,
  point_size = 1
```

```
giotto object
gobject
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
dim1_to_use
                 dimension to use on x-axis
                 dimension to use on y-axis
dim2_to_use
dim3_to_use
                 dimension to use on z-axis
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
```

276 visForceLayoutPlot

```
name of NN network to use, if show_NN_network = TRUE
network_name
color_as_factor
                  convert color column to factor
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
                  column to use for alpha of the edges
edge_alpha
point_size
                  size of point (cell)
```

### **Details**

Description of parameters.

### Value

plotly

# **Examples**

```
visDimPlot_3D_plotly(gobject)
```

visForceLayoutPlot visForceLayoutPlot

# Description

Visualize cells according to forced layout algorithm coordinates

```
visForceLayoutPlot(
  gobject,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  layout_name = "layout",
  dim1_to_use = 1,
  dim2_to_use = 2,
  show_NN_network = T,
  cell_color = NULL,
  color_as_factor = TRUE,
  cell_color_code = NULL,
  edge_alpha = NULL,
  point_size = 1,
```

visForceLayoutPlot 277

```
point_border_col = "black",
point_border_stroke = 0.1,
show_legend = T,
show_plot = F,
return_plot = TRUE,
save_plot = F,
save_dir = NULL,
save_folder = NULL,
save_name = NULL,
save_format = NULL,
show_saved_plot = F,
...
)
```

```
gobject
                  giotto object
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
network_name
                  NN network to use
                  name of layout to use
layout_name
dim1_to_use
                  dimension to use on x-axis
                  dimension to use on y-axis
dim2_to_use
show_NN_network
                  show underlying NN network
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
                  column to use for alpha of the edges
edge_alpha
                  size of point (cell)
point_size
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
                  show legend
show_legend
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  directory to save the plot
save_dir
save_folder
                  (optional) folder in directory to save the plot
                  name of plot
save_name
save_format
                  format of plot (e.g. tiff, png, pdf, ...)
show_saved_plot
                  load & display the saved plot
```

278 visGenePlot

#### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

visForceLayoutPlot(gobject)

visGenePlot

visGenePlot

# **Description**

Visualize cells and gene expression according to spatial coordinates

```
visGenePlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes_high_color = NULL,
  genes_mid_color = "white",
  genes_low_color = "blue",
  show_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  edge_alpha = NULL,
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  midpoint = 0,
  scale_alpha_with_expression = FALSE,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
  plot_method = c("ggplot", "plotly"),
  show_plots = F
```

visGenePlot 279

### **Arguments**

gobject giotto object

expression\_values

gene expression values to use

genes genes to show

genes\_high\_color

color represents high gene expression

genes\_mid\_color

color represents middle gene expression

genes\_low\_color

color represents low gene expression

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

show\_grid show spatial grid grid\_color color of spatial grid

spatial\_grid\_name

name of spatial grid to use

midpoint expression midpoint
scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

show\_legend show legend

cow\_n\_col cowplot param: how many columns cow\_rel\_h cowplot param: relative height cow\_rel\_w cowplot param: relative width cow\_align cowplot param: how to align three mode to adjust axis scale axis\_scale x\_ticks number of ticks on x axis number of ticks on y axis y\_ticks number of ticks on z axis z\_ticks plot\_method two methods of plot

show plots

# **Details**

Description of parameters.

### Value

ggplot or plotly

show\_plots

### **Examples**

```
visGenePlot(gobject)
```

```
visGenePlot_2D_ggplot visGenePlot_2D_ggplot
```

# **Description**

Visualize cells and gene expression according to spatial coordinates

### Usage

```
visGenePlot_2D_ggplot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  genes_high_color = "darkred",
  genes_mid_color = "white",
  genes_low_color = "darkblue",
  show_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  edge_alpha = NULL,
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  midpoint = 0,
  scale_alpha_with_expression = FALSE,
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  show_plots = F
)
```

```
gobject giotto object
expression_values
gene expression values to use
genes genes to show
genes_high_color
color represents high gene expression
genes_mid_color
color represents middle gene expression
```

visGenePlot\_3D\_plotly

```
genes_low_color
```

color represents low gene expression

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

show\_grid show spatial grid grid\_color color of spatial grid

spatial\_grid\_name

name of spatial grid to use

midpoint expression midpoint

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

show\_legend show legend

cow\_n\_colcowplot param: how many columnscow\_rel\_hcowplot param: relative heightcow\_rel\_wcowplot param: relative widthcow\_aligncowplot param: how to align

show\_plots show plots

# **Details**

Description of parameters.

# Value

ggplot

# **Examples**

visGenePlot\_2D\_ggplot(gobject)

 ${\tt visGenePlot\_3D\_plotly} \ \ {\it visGenePlot\_3D\_plotly}$ 

# Description

Visualize cells and gene expression according to spatial coordinates

### Usage

```
visGenePlot_3D_plotly(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  show_network = F,
  network_color = NULL,
  spatial_network_name = "spatial_network",
  edge_alpha = NULL,
  show\_grid = F,
  genes_high_color = NULL,
  genes_mid_color = "white",
  genes_low_color = "blue",
  spatial_grid_name = "spatial_grid",
  point_size = 1,
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
  show_plots = F
gobject
                giotto object
expression_values
                gene expression values to use
```

```
genes
                  genes to show
                  show underlying spatial network
show_network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
show_grid
                  show spatial grid
genes_high_color
                  color represents high gene expression
genes_mid_color
                  color represents middle gene expression
genes_low_color
                  color represents low gene expression
spatial_grid_name
                  name of spatial grid to use
                  size of point (cell)
point_size
show_legend
                  show legend
axis_scale
                  three mode to adjust axis scale
x_ticks
                  number of ticks on x axis
                  number of ticks on y axis
y_ticks
```

visPlot 283

### **Details**

Description of parameters.

#### Value

plotly

# **Examples**

```
visGenePlot_3D_plotly(gobject)
```

visPlot visPlot

# Description

Visualize cells according to spatial coordinates

```
visPlot(
  gobject,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  point_size = 3,
  point_border_col = "black",
  point_border_stroke = 0.1,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  show_network = F,
  network_color = NULL,
  network_alpha = 1,
  other_cell_alpha = 0.1,
  spatial_network_name = "spatial_network",
  show\_grid = F,
```

284 visPlot

```
grid_color = NULL,
 grid_alpha = 1,
  spatial_grid_name = "spatial_grid",
  coord_fix_ratio = 0.6,
  title = "",
  show_legend = T,
 axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_{ticks} = NULL,
 y_ticks = NULL,
  z_ticks = NULL,
 plot_method = c("ggplot", "plotly"),
  show_plot = F,
  return_plot = TRUE,
  save_plot = F,
  save_dir = NULL,
  save_folder = NULL,
  save_name = NULL,
  save_format = NULL,
 show_saved_plot = F,
)
```

```
gobject
                  giotto object
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
sdimz
                  z-axis dimension name (default = 'sdimz')
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
show_network
                  show underlying spatial network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
```

visPlot\_2D\_ggplot 285

```
show_grid
                  show spatial grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
coord_fix_ratio
                  fix ratio between x and y-axis
title
                  title of plot
show_legend
                  show legend
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_dir
                  directory to save the plot
                  (optional) folder in directory to save the plot
save_folder
                  name of plot
save_name
save_format
                  format of plot (e.g. tiff, png, pdf, ...)
show_saved_plot
                  load & display the saved plot
```

#### **Details**

Description of parameters.

### Value

ggplot

# **Examples**

```
visPlot(gobject)
```

# **Description**

Visualize cells according to spatial coordinates

```
visPlot_2D_ggplot(
  gobject,
  sdimx = NULL,
  sdimy = NULL,
  point_size = 3,
  point_border_col = "black",
  point_border_stroke = 0.1,
  cell_color = NULL,
  cell_color_code = NULL,
```

286 visPlot\_2D\_ggplot

```
color_as_factor = T,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  show_network = F,
  network_color = NULL,
  network_alpha = 1,
  other_cells_alpha = 0.1,
  spatial_network_name = "spatial_network",
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  coord_fix_ratio = 0.6,
  title = "",
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_{ticks} = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
  show_plot = F,
  return_plot = TRUE,
  save_plot = F,
  save_dir = NULL,
  save_folder = NULL,
  save_name = NULL,
  save_format = NULL,
  show_saved_plot = F,
)
```

```
gobject
                  giotto object
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
color_as_factor
                  convert color column to factor
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
```

visPlot\_2D\_ggplot 287

show\_other\_cells

display not selected cells

other\_cell\_color

color of not selected cells

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

show\_grid show spatial grid

grid\_color color of spatial grid

spatial\_grid\_name

name of spatial grid to use

coord\_fix\_ratio

fix ratio between x and y-axis

title title of plot

show\_legend show legend

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

save\_dir directory to save the plot

save\_folder (optional) folder in directory to save the plot

save\_name name of plot

save\_format format of plot (e.g. tiff, png, pdf, ...)

show\_saved\_plot

load & display the saved plot

# **Details**

Description of parameters.

### Value

ggplot

### **Examples**

 ${\tt visPlot\_2D\_ggplot(gobject)}$ 

288 visPlot\_2D\_plotly

```
visPlot_2D_plotly visPlot_2D_plotly
```

# **Description**

Visualize cells according to spatial coordinates

#### Usage

```
visPlot_2D_plotly(
  gobject,
  sdimx = NULL,
  sdimy = NULL,
  point_size = 3,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_network = F,
  network_color = "lightgray",
  network_alpha = 1,
  other_cell_alpha = 0.5,
  spatial_network_name = "spatial_network",
  show\_grid = F,
  grid_color = NULL,
  grid_alpha = 1,
  spatial_grid_name = "spatial_grid",
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  show_plot = F
```

```
gobject giotto object

sdimx x-axis dimension name (default = 'sdimx')

sdimy y-axis dimension name (default = 'sdimy')

point_size size of point (cell)

cell_color color for cells (see details)

cell_color_code

named vector with colors

color_as_factor

convert color column to factor
```

visPlot\_3D\_plotly 289

```
select_cell_groups
                  select a subset of the groups from cell_color
                  show underlying spatial network
show_network
                  color of spatial network
network_color
spatial_network_name
                  name of spatial network to use
                  show spatial grid
show_grid
grid_color
                  color of spatial grid
                  alpha of spatial grid
grid_alpha
spatial_grid_name
                  name of spatial grid to use
                  show legend
show_legend
show_plot
                  show plot
```

#### **Details**

Description of parameters.

#### Value

plotly

# **Examples**

```
visPlot_2D_plotly(gobject)
```

```
visPlot_3D_plotly
```

# Description

Visualize cells according to spatial coordinates

```
visPlot_3D_plotly(
  gobject,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  point_size = 3,
  cell_color = NULL,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_network = F,
```

290 visPlot\_3D\_plotly

```
network_color = NULL,
network_alpha = 1,
other_cell_alpha = 0.5,
spatial_network_name = "spatial_network",
spatial_grid_name = "spatial_grid",
title = "",
show_legend = T,
axis_scale = c("cube", "real", "custom"),
custom_ratio = NULL,
x_ticks = NULL,
y_ticks = NULL,
z_ticks = NULL,
show_plot = F
)
```

# Arguments

```
gobject
                  giotto object
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
sdimz
                  z-axis dimension name (default = 'sdimz')
point_size
                  size of point (cell)
cell_color
                  color for cells (see details)
cell_color_code
                  named vector with colors
select_cell_groups
                  select a subset of the groups from cell_color
                  show underlying spatial network
show_network
network_color
                  color of spatial network
spatial_network_name
                  name of spatial network to use
spatial_grid_name
                  name of spatial grid to use
                  title of plot
title
show_legend
                  show legend
show_plot
                  show plot
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
color_as_factor
                  convert color column to factor
show_grid
                  show spatial grid
grid_color
                  color of spatial grid
coord_fix_ratio
                  fix ratio between x and y-axis
```

visSpatDimGenePlot 291

#### **Details**

Description of parameters.

#### Value

ggplot

#### **Examples**

```
visPlot_3D_plotly(gobject)
```

visSpatDimGenePlot

visSpatDimGenePlot

## **Description**

integration of visSpatDimGenePlot\_2D(ggplot) and visSpatDimGenePlot\_3D(plotly)

```
visSpatDimGenePlot(
 gobject,
 plot_method = c("ggplot", "plotly"),
 expression_values = c("normalized", "scaled", "custom"),
 plot_alignment = c("horizontal", "vertical"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
 dim3_to_use = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
 genes,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 edge_alpha_dim = NULL,
  scale_alpha_with_expression = FALSE,
 label_size = 16,
 genes_low_color = "blue",
 genes_mid_color = "white",
 genes_high_color = "red",
 dim_point_size = 3,
 nn_network_alpha = 0.5,
  show_spatial_network = F,
  spatial_network_name = "spatial_network",
 network_color = "lightgray",
  spatial_network_alpha = 0.5,
```

292 visSpatDimGenePlot

```
show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_grid_alpha = 0.5,
      spatial_point_size = 3,
      spatial_point_border_col = "black",
      spatial_point_border_stroke = 0.1,
      legend_text_size = 12,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
     x_ticks = NULL,
     y_ticks = NULL,
     z_ticks = NULL,
     midpoint = 0,
     point_size = 1,
      cow_n_col = 2,
      cow_rel_h = 1,
     cow_rel_w = 1,
     cow_align = "h",
     show_legend = T,
      show_plots = F
   )
Arguments
   gobject
                    giotto object
   expression_values
                    gene expression values to use
   plot_alignment direction to align plot
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
   dim1_to_use
                    dimension to use on x-axis
   dim2_to_use
                    dimension to use on y-axis
   dim3_to_use
                    dimension to use on z-axis
   sdimx
                    x-axis dimension name (default = 'sdimx')
   sdimy
                    y-axis dimension name (default = 'sdimy')
    sdimz
                    z-axis dimension name (default = 'sdimz')
   genes
                    genes to show
   dim_point_border_col
                    color of border around points
   dim_point_border_stroke
                    stroke size of border around points
   show_NN_network
                    show underlying NN network
   nn_network_to_use
                    type of NN network to use (kNN vs sNN)
                    name of NN network to use, if show_NN_network = TRUE
   network_name
```

edge\_alpha\_dim dim reduction plot: column to use for alpha of the edges scale\_alpha\_with\_expression scale expression with ggplot alpha parameter

label\_size size for the label

genes\_low\_color

color to represent low expression of gene

genes\_high\_color

color to represent high expression of gene

dim\_point\_size dim reduction plot: point size

spatial\_network\_name

name of spatial network to use

spatial\_grid\_name

name of spatial grid to use

spatial\_point\_size

spatial plot: point size

spatial\_point\_border\_col

color of border around points

spatial\_point\_border\_stroke

stroke size of border around points

legend\_text\_size

the size of the text in legend

axis\_scale three modes to adjust axis scale ratio custom\_ratio set the axis scale ratio on custom

 $x_{ticks}$  number of ticks on x axis  $y_{ticks}$  number of ticks on y axis  $z_{ticks}$  number of ticks on z axis

midpoint size of point (cell)
point\_size size of point (cell)

cow\_n\_col cowplot param: how many columns
cow\_rel\_h cowplot param: relative height
cow\_rel\_w cowplot param: relative width
cow\_align cowplot param: how to align

show\_legend show legend
show\_plot show plot

#### **Details**

Description of parameters.

#### Value

ggplot or plotly

## **Examples**

 $\verb|visSpatDimGenePlot(gobject)| \\$ 

visSpatDimGenePlot\_2D visSpatDimGenePlot\_2D

## **Description**

Visualize cells according to spatial AND dimension reduction coordinates in ggplot mode

## Usage

```
visSpatDimGenePlot_2D(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  plot_alignment = c("horizontal", "vertical"),
  genes,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  point_size = 1,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  show_NN_network = F,
  show_spatial_network = F,
  show_spatial_grid = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  edge_alpha_dim = NULL,
  scale_alpha_with_expression = FALSE,
  spatial_network_name = "spatial_network",
  spatial_grid_name = "spatial_grid",
  spatial_point_size = 1,
  spatial_point_border_col = "black",
  spatial_point_border_stroke = 0.1,
  midpoint = 0,
  genes_high_color = "red",
  genes_mid_color = "white";
  genes_low_color = "blue",
  cow_n_col = 2,
  cow_rel_h = 1,
  cow_rel_w = 1,
  cow_align = "h",
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  show_legend = T,
  show_plots = F
```

#### **Arguments**

gobject giotto object

expression\_values

gene expression values to use

plot\_alignment direction to align plot

genes genes to show

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimension reduction name

dim1\_to\_use dimension to use on x-axis dim2\_to\_use dimension to use on y-axis

point\_size size of point (cell)

dim\_point\_border\_col

color of border around points

dim\_point\_border\_stroke

stroke size of border around points

show\_NN\_network

show underlying NN network

 $nn\_network\_to\_use$ 

type of NN network to use (kNN vs sNN)

 $network\_name \qquad name \ of \ NN \ network \ to \ use, \ if \ show\_NN\_network = TRUE$ 

edge\_alpha\_dim dim reduction plot: column to use for alpha of the edges

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

spatial\_network\_name

name of spatial network to use

spatial\_grid\_name

name of spatial grid to use

spatial\_point\_size

spatial plot: point size

spatial\_point\_border\_col

color of border around points

spatial\_point\_border\_stroke

stroke size of border around points

midpoint size of point (cell)

cow\_n\_col cowplot param: how many columns

cow\_rel\_h cowplot param: relative height cow\_rel\_w cowplot param: relative width cow\_align cowplot param: how to align

show\_legend show legend

dim\_point\_size dim reduction plot: point size

show\_plot show plot

#### **Details**

Description of parameters.

#### Value

ggplot

#### **Examples**

```
visSpatDimGenePlot_2D(gobject)
```

```
visSpatDimGenePlot_3D visSpatDimGenePlot_3D
```

## **Description**

Visualize cells according to spatial AND dimension reduction coordinates in plotly mode

```
visSpatDimGenePlot_3D(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  plot_alignment = c("horizontal", "vertical"),
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  dim3_to_use = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  genes,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  label_size = 16,
  genes_low_color = "blue",
  genes_mid_color = "white",
  genes_high_color = "red",
  dim_point_size = 3,
  nn_network_alpha = 0.5,
  show\_spatial\_network = F,
  spatial_network_name = "spatial_network",
  network_color = "lightgray",
  spatial_network_alpha = 0.5,
  show_spatial_grid = F,
  spatial_grid_name = "spatial_grid",
  spatial_grid_color = NULL,
  spatial_grid_alpha = 0.5,
  spatial_point_size = 3,
  legend_text_size = 12,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_{ticks} = NULL,
```

y\_ticks = NULL, z\_ticks = NULL

```
Arguments
   gobject
                     giotto object
    plot_alignment direction to align plot
   dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
   dim1_to_use
   dim2_to_use
                     dimension to use on y-axis
    dim3_to_use
                     dimension to use on z-axis
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
   network_name
    genes_low_color
                     color represent high gene expression (see details)
    genes_high_color
                     color represent high gene expression (see details)
    nn_network_alpha
                     column to use for alpha of the edges
    show_spatial_network
                     show spatial network
    spatial_network_name
                     name of spatial network to use
    network_color color of spatial/nn network
    spatial_network_alpha
                     alpha of spatial network
    show_spatial_grid
                     show spatial grid
    spatial_grid_name
                     name of spatial grid to use
    spatial_grid_color
                     color of spatial grid
    spatial_grid_alpha
                     alpha of spatial grid
```

## **Details**

Description of parameters.

text size of legend

show legend show plot

legend\_text\_size

show\_legend

show\_plot

298 visSpatDimPlot

#### Value

plotly

#### **Examples**

```
visSpatDimPlot_3D(gobject)
```

visSpatDimPlot

visSpatDimPlot

## **Description**

integration of visSpatDimPlot\_2D and visSpatDimPlot\_3D

```
visSpatDimPlot(
  gobject,
  plot_method = c("ggplot", "plotly"),
  plot_alignment = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  dim3_to_use = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = NULL,
  label_fontface = "bold",
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  dim_point_size = 3,
  dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  nn_network_alpha = NULL,
  show_spatial_network = F,
  spatial_network_name = "spatial_network",
  network_color = "lightgray",
  spatial_network_alpha = 0.5,
```

visSpatDimPlot 299

```
show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_grid_alpha = 0.5,
      spatial_point_size = 3,
      legend_text_size = 12,
      spatial_point_border_col = "black",
      spatial_point_border_stroke = 0.1,
      show_legend = T,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
      x_ticks = NULL,
      y_ticks = NULL,
      z_ticks = NULL,
      show_plot = F
Arguments
    gobject
                     giotto object
    plot_alignment direction to align plot
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    dim3_to_use
                     dimension to use on z-axis
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
                     select subset of cells based on cell IDs
    select_cells
    show_other_cells
                     display not selected cells
    other_cell_color
                     color of not selected cells
    nn_network_alpha
                     column to use for alpha of the edges
    show\_spatial\_network
                     show spatial network
```

300 visSpatDimPlot\_2D

```
spatial_network_name
                 name of spatial network to use
spatial_network_alpha
                 alpha of spatial network
show_spatial_grid
                 show spatial grid
spatial_grid_name
                 name of spatial grid to use
spatial_grid_color
                 color of spatial grid
spatial_grid_alpha
                 alpha of spatial grid
legend_text_size
                 text size of legend
show_legend
                 show legend
show_plot
                 show plot
plot_mode
                 choose the mode to draw plot: ggplot or plotly
spatial_network_color
                 color of spatial network
```

#### **Details**

Description of parameters.

#### Value

ggplot or plotly

#### **Examples**

```
visSpatDimPlot(gobject)
```

visSpatDimPlot\_2D
visSpatDimPlot\_2D

## **Description**

Visualize cells according to spatial AND dimension reduction coordinates in ggplot2 mode

```
visSpatDimPlot_2D(
  gobject,
  plot_alignment = c("vertical", "horizontal"),
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2_to_use = 2,
  sdimx = NULL,
  sdimy = NULL,
```

visSpatDimPlot\_2D

 $show_NN_network = F,$ 

nn\_network\_to\_use

network\_name
cell\_color

301

```
nn_network_to_use = "sNN",
     network_name = "sNN.pca",
      show\_cluster\_center = F,
      show_center_label = T,
      center_point_size = 4,
      label_size = 4,
      label_fontface = "bold",
      cell_color = NULL,
      color_as_factor = T,
      cell_color_code = NULL,
      select_cell_groups = NULL,
      select_cells = NULL,
      show_other_cells = T,
     other_cell_color = "lightgrey",
      dim_plot_mode = NULL,
     dim_point_size = 1,
     dim_point_border_col = "black",
     dim_point_border_stroke = 0.1,
     nn_network_alpha = 0.05,
      show_spatial_network = F,
      spatial_network_name = "spatial_network",
      spatial_network_color = NULL,
      show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_point_size = 1,
      spatial_point_border_col = "black",
      spatial_point_border_stroke = 0.1,
      show_legend = T,
      show_plot = F,
     plot_method = "ggplot"
Arguments
                    giotto object
   gobject
   plot_alignment direction to align plot
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
   dim1_to_use
                    dimension to use on x-axis
                    dimension to use on y-axis
   dim2_to_use
   show_NN_network
                    show underlying NN network
```

type of NN network to use (kNN vs sNN)

color for cells (see details)

name of NN network to use, if show\_NN\_network = TRUE

302 visSpatDimPlot\_2D

```
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
nn_network_alpha
                  column to use for alpha of the edges
show_spatial_network
                  show spatial network
spatial_network_name
                  name of spatial network to use
spatial_network_color
                  color of spatial network
show_spatial_grid
                  show spatial grid
spatial_grid_name
                  name of spatial grid to use
spatial_grid_color
                  color of spatial grid
show_legend
                  show legend
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
save_dir
                  directory to save the plot
                  (optional) folder in directory to save the plot
save_folder
                  name of plot
save_name
save_format
                  format of plot (e.g. tiff, png, pdf, ...)
show_saved_plot
                  load & display the saved plot
```

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

```
visSpatDimPlot_2D(gobject)
```

visSpatDimPlot\_3D 303

visSpatDimPlot\_3D 1

visSpatDimPlot\_3D

# Description

Visualize cells according to spatial AND dimension reduction coordinates in plotly mode

## Usage

```
visSpatDimPlot_3D(
  gobject,
  plot_alignment = c("horizontal", "vertical"),
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  dim3_to_use = NULL,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = 16,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  dim_point_size = 3,
  nn_network_alpha = 0.5,
  show_spatial_network = F,
  spatial_network_name = "spatial_network",
  network_color = "lightgray",
  spatial_network_alpha = 0.5,
  show_spatial_grid = F,
  spatial_grid_name = "spatial_grid",
  spatial_grid_color = NULL,
  spatial_grid_alpha = 0.5,
  spatial_point_size = 3,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
  legend_text_size = 12
```

#### **Arguments**

gobject giotto object

304 visSpatDimPlot\_3D

```
plot_alignment direction to align plot
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
                 dimension to use on x-axis
dim1_to_use
                 dimension to use on y-axis
dim2_to_use
dim3_to_use
                 dimension to use on z-axis
show_NN_network
                 show underlying NN network
nn_network_to_use
                 type of NN network to use (kNN vs sNN)
                 name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                 color for cells (see details)
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
nn_network_alpha
                 column to use for alpha of the edges
show_spatial_network
                 show spatial network
spatial_network_name
                 name of spatial network to use
spatial_network_alpha
                 alpha of spatial network
show_spatial_grid
                 show spatial grid
spatial_grid_name
                 name of spatial grid to use
spatial_grid_color
                 color of spatial grid
spatial_grid_alpha
                 alpha of spatial grid
legend_text_size
                 text size of legend
spatial_network_color
                 color of spatial network
                 show legend
show_legend
                 show plot
show_plot
```

## **Details**

Description of parameters.

#### Value

plotly

writeHMRFresults 305

## **Examples**

```
visSpatDimPlot_3D(gobject)
```

writeHMRFresults

writeHMRFresults

#### **Description**

write results from doHMRF to a data.table.

#### Usage

```
writeHMRFresults(
  gobject,
  HMRFoutput,
  k = NULL,
  betas_to_view = NULL,
  print_command = F
)
```

# Arguments

gobject giotto object

HMRF output HMRF output from doHMRF

k k to write results for

betas\_to\_view results from different betas that you want to view

print\_command see the python command

# Value

data.table with HMRF results for each b and the selected k

# **Examples**

```
writeHMRFresults(gobject)
```

## **Description**

write out factor-like annotation data from a giotto object for the Viewer

```
write_giotto_viewer_annotation(
  annotation,
  annot_name = "test",
  output_directory = getwd()
)
```

#### **Arguments**

```
annotation annotation from the data.table from giotto object
annot_name name of the annotation
output_directory
directory where to save the files
```

#### Value

write a .txt and .annot file for the selection annotation

#### **Description**

write out dimensional reduction data from a giotto object for the Viewer

## Usage

```
write_giotto_viewer_dim_reduction(
  dim_reduction_cell,
  dim_red = NULL,
  dim_red_name = NULL,
  dim_red_rounding = NULL,
  dim_red_rescale = c(-20, 20),
  output_directory = getwd()
)
```

## **Arguments**

```
dim_reduction_cell

dimension reduction slot from giotto object

dim_red high level name of dimension reduction

dim_red_name specific name of dimension reduction to use

dim_red_rounding

numerical indicating how to round the coordinates

dim_red_rescale

numericals to rescale the coordinates

output_directory

directory where to save the files
```

#### Value

write a .txt and .annot file for the selection annotation

# Description

write out numeric annotation data from a giotto object for the Viewer

# Usage

```
write_giotto_viewer_numeric_annotation(
  annotation,
  annot_name = "test",
  output_directory = getwd()
)
```

# **Arguments**

```
annotation annotation from the data.table from giotto object
annot_name name of the annotation
output_directory
directory where to save the files
```

# Value

write a .txt and .annot file for the selection annotation

# Index

Tania giotto	aallDuavimitullaatuun 26
*Topic <b>giotto</b> ,	cellProximityHeatmap, 26
giotto-class, 137	cellProximityNetwork, 27
print.giotto, 180	cellProximitySpatPlot, 29
show, giotto-method, 189	cellProximitySpatPlot2D, 30, 30
*Topic <b>giotto</b>	cellProximitySpatPlot3D, 30, 32
createGiottoObject, 47	cellProximityVisPlot, 34
*Topic <b>object</b>	cellProximityVisPlot_2D_ggplot, 36
giotto-class, 137	cellProximityVisPlot_2D_plotly, 38
print.giotto, 180	<pre>cellProximityVisPlot_3D_plotly, 40</pre>
show,giotto-method,189	changeGiottoInstructions, 41
	cluster_walktrap, 102
addCellMetadata, 7, 49	clusterCells, 42
addCellStatistics, 7, 12	clusterSpatialCorGenes, 45
addGeneMetadata, $8,49$	combineMetadata, 45
addGeneStatistics, 9, <i>12</i>	<pre>convertEnsemblToGeneSymbol, 46</pre>
addHMRF, 10	cowplot::save_plot, 137
addNetworkLayout, 11	<pre>create_average_detection_DT, 58</pre>
addStatistics, 12	create_average_DT, 58
adjustGiottoMatrix, 12	<pre>create_cell_type_random_cell_IDs, 59</pre>
aes_string2, 13	create_cluster_matrix, 60
all_plots_save_function, 15, 21, 25, 27,	<pre>create_dimObject,60</pre>
28, 30, 32, 34, 66, 69, 71, 73, 74, 77,	createGiottoInstructions, $46,48$
80, 83, 85, 139, 150, 152, 156, 158,	createGiottoObject, 47, 257
160, 162, 164, 166, 168, 169, 171,	createHeatmap_DT,49
173, 175, 182, 190, 191, 193, 196,	createMetagenes, 50
198, 199, 206, 209, 213, 217, 219,	createNearestNetwork, 51
222, 224, 228, 232, 235, 237, 239,	createSpatialEnrich, 53, 108
241, 246, 249, 252, 254, 264	createSpatialGrid, 54
allCellCellcommunicationsScores, 14	createSpatialGrid_2D,55
<pre>annotate_spatlocs_with_spatgrid_2D, 17</pre>	createSpatialGrid_3D, 56
<pre>annotate_spatlocs_with_spatgrid_3D, 18</pre>	createSpatialNetwork, 57
annotateGiotto, 16	
annotateSpatialNetwork, 17	decide_cluster_order,61
<pre>average_gene_gene_expression_in_groups,</pre>	detectSpatialCorGenes, 62
18	detectSpatialPatterns, 63
	dimCellPlot, 64
binGetSpatialGenes, 19	dimCellPlot2D, 66, 67
	dimGenePlot, 69
<pre>calculate_spatial_genes_python, 23</pre>	dimGenePlot2D, 71
calculateHVG, 20	dimGenePlot3D, <i>71</i> , <i>73</i> , <i>73</i>
calculateMetaTable, 22	dimPlot, 75
calculateMetaTableCells, 23	dimPlot2D, 77, 78, 160, 162, 166, 168, 171,
cellProximityBarplot, 24	173
cellProximityEnrichment, 25	dimPlot2D_single, 81
	= 5 /-

INDEX 309

dimPlot3D, 69, 77, 80, 83, 83	get10Xmatrix, 129
direction_test (direction_test_CPG), 85	<pre>get_cell_to_cell_sorted_name_conversion,</pre>
direction_test_CPG, 85	134
do_spatial_grid_averaging, 104	<pre>get_interaction_gene_enrichment, 135</pre>
do_spatial_knn_smoothing, 105	<pre>get_specific_interaction_gene_enrichment,</pre>
doHclust, 44, 86	135
doHMRF, 87	<pre>getCellProximityGeneScores, 129</pre>
doKmeans, 44, 88	getClusterSimilarity, 131
doLeidenCluster, 44, 90, 93, 258	<pre>getDendrogramSplits, 132</pre>
doLeidenSubCluster, 91	<pre>getDistinctColors, 133</pre>
doLouvainCluster, 44, 93	getGeneToGeneScores, 133
doLouvainCluster_community, 44, 94, 94,	<pre>ggplot_save_function, 136</pre>
98, 99, 258	giotto (giotto-class), 137
doLouvainCluster_multinet, 44, 94, 95, 98,	giotto-class, 137
101, 258	glouvain_ml, 96
doLouvainSubCluster, 96	
doLouvainSubCluster_community, 98	hclust, <i>87</i>
doLouvainSubCluster_multinet, 100	Heatmap, <i>139</i>
doRandomWalkCluster, 44, 101	heatmSpatialCorGenes, 138
doSNNCluster, 44, 103	hyperGeometricEnrich, 53, 139
dt_to_matrix, 106	
/	kmeans, $89$
enrichSpatialCorGroups, 106	kmeans_binarize, 140
exportGiottoViewer, 107	kNN, <i>52</i>
exprOnlyCellCellcommunicationScores,	1
108	layout_with_drl, 11
extended_gini_fun, 109	loadHMRF, 140
extractNearestNetwork, 110	make eimuleted network 142
,	make_simulated_network, 142
fDataDT, 110	makeSignMatrixPAGE, 141, 147
filterCombinations, 111, 114	makeSignMatrixRank, 141, 181
filterCPGscores, 112	mergeClusters, 142
filterDistributions, 113	mygini_fun, 144
filterGiotto, 114	nnDT_to_kNN, 144
find_grid_2D, 124	node_clusters, 144
find_grid_3D, 124	normalizeGiotto, 145
find_grid_x, 124	normalizedicto, 143
find_grid_y, 125	OR_function2, 146
find_grid_z, 125	,
findGiniMarkers, 115, 117, 118	PAGEEnrich, 53, 141, 147
findGiniMarkers_one_vs_all, 116, 120	PCA, <i>185</i>
findMarkers, 117, 123	pDataDT, 148
findMarkers_one_vs_all, 119	permutationPA, 204
findMastMarkers, 118, 120, 122	plot_network_layer_ggplot, 175
findMastMarkers_one_vs_all, 120, 121	plot_point_layer_ggplot, 176
findScranMarkers, 118, 122, 124	plot_spat_point_layer_ggplot, 178
findScranMarkers_one_vs_all, 120, 123	plotCPGscores, 148
fish_function, 125	plotGTGscores, 149
fish_function2, 125	plotHeatmap, 151
FSV_show, 126	plotly_axis_scale_2D, 153
_ ,	plotly_axis_scale_3D, 153
GenePattern_show, 127	plotly_grid, 154
general_save_function, 16, 128	plotly_network, 155

310 INDEX

plotMetaDataCellsHeatmap, 155, 159	spatDimPlot2D, 228, 228
plotMetaDataHeatmap, 157, 157	spatDimPlot3D, 228, 232, 232
plotPCA, 159	spatGenePlot, 236
plotPCA_2D, 161	spatGenePlot2D, 237, 238
plotPCA_3D, 160, 162, 163	spatGenePlot3D, 237, 239, 240
plotTSNE, 164	Spatial_AEH, 241
plotTSNE_2D, 166	Spatial_DE, 242
plotTSNE_3D, 166, 168, 168	spatPlot, 243
plotUMAP, 170	spatPlot2D, 246, 261
plotUMAP_2D, 172	spatPlot2D_single, 249
plotUMAP_3D, <i>171</i> , <i>173</i> , 174	spatPlot3D, 246, 249, 252, 252, 262
print.giotto, 180	specificCellCellcommunicationScores,
	254
rank_binarize, 182	<pre>split_dendrogram_in_two, 255</pre>
rankEnrich, <i>53</i> , <i>142</i> , 180	stitchFieldCoordinates, 49, 256
rankSpatialCorGroups, 181	subClusterCells, 257
readGiottoInstructions, 182	subsetGiotto, 259
removeCellAnnotation, 183	subsetGiottoLocs, 259
removeGeneAnnotation, 183	
replaceGiottoInstructions, 184	umap, 188
Rtsne, <i>186</i>	viewHMRFresults, 260
runPCA, 184	viewHMRFresults2D, 261
runtSNE, 185	viewHMRFresults3D, 262
runUMAP, 187	violinPlot, 263
	visDimGenePlot, 264
select_expression_values, 189	
selectPatternGenes, 188	visDimGenePlot_2D_ggplot, 266
show, giotto-method, 189	visDimGenePlot_3D_plotly, 267
showClusterDendrogram, 190	visDimPlot, 269
showClusterHeatmap, 191	visDimPlot_2D_ggplot, 271
showCPGscores, 192	visDimPlot_2D_plotly, 273
showGeneExpressionProximityScore, 193	visDimPlot_3D_plotly, 275
showGiottoInstructions, 194	visForceLayoutPlot, 276
showGTGscores, 194	visGenePlot, 278
showIntExpressionProximityScore, 195	visGenePlot_2D_ggplot, 280
showPattern, 196	visGenePlot_3D_plotly, 281
showPattern2D, <i>197</i> , 197	visPlot, 261, 283
showPattern3D, 198	visPlot_2D_ggplot, 285
showPatternGenes, 199	visPlot_2D_plotly, 288
showProcessingSteps, 200	visPlot_3D_plotly, 289
showSpatialCorGenes, 63, 201	visSpatDimGenePlot, 291
showTopGeneToGene, 202	visSpatDimGenePlot_2D, 294
signPCA, 203	visSpatDimGenePlot_3D, 296
sNN, 52	visSpatDimPlot, 298
sNNclust, 103	visSpatDimPlot_2D, 300
spatCellPlot, 204	visSpatDimPlot_3D, 303
spatCellPlot2D, 207	write_giotto_viewer_annotation, 305
spatDimCellPlot, 209	write_giotto_viewer_dim_reduction, 306
spatDimCellPlot2D, 213	write_giotto_viewer_numeric_annotation,
spatDimGenePlot, 217	307
spatDimGenePlot2D, 220	writeHMRFresults, 305
spatDimGenePlot3D, 220, 222, 222	250111111 1 5502 55, 505
spatDimPlot, 224	zlm, <i>121</i>