# Package 'Giotto'

February 7, 2020

```
Title Spatial single-cell transcriptomics pipeline.
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      ggplot2 (>= 3.1.1),
      base (>= 3.5.1),
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      R (>= 3.5.1)
Imports Rtsne (>= 0.15),
      uwot (>= 0.0.0.9010),
      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,
      ggraph,
```

2 R topics documented:

```
Rcpp,
     rlang (>= 0.4.3)
Suggests knitr,
     rmarkdown,
     MAST,
     scran (>= 1.10.1),
     png,
     tiff,
     biomaRt,
     trendsceek,
     multinet (>= 3.0.2)
biocViews
VignetteBuilder knitr
LinkingTo Rcpp,
     RcppArmadillo
Remotes lambdamoses/smfishhmrf-r
```

# **R** topics documented:

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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

8 addCellStatistics

# **Examples**

```
addCellMetadata(gobject)
```

addCellStatistics

addCellStatistics

# **Description**

adds cells statistics to the giotto object

# 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 9

addGeneMetadata

addGeneMetadata

## **Description**

adds gene metadata to the giotto object

## Usage

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

# **Arguments**

```
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

#### **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

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

10 addHMRF

#### **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

#### Value

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

#### **Examples**

```
addGeneStatistics(gobject)
```

addHMRF

# Description

Add selected results from doHMRF to the giotto object

addHMRF

# Usage

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

# Arguments

gobject giotto object

HMRF output from doHMRF()

k number of domains

betas\_to\_add results from different betas that you want to add

name specify a custom name

# **Details**

Description ...

addNetworkLayout 11

#### Value

```
giotto object
```

### **Examples**

```
addHMRF(gobject)
```

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 <a href="layout\_with\_drl">layout\_with\_drl</a>, 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,
)
```

# Arguments

```
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
```

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```
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)
```

# 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_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
)
```

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# **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

```
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 = TRUE,
  get_av_expr = TRUE,
  get_high_expr = TRUE,
  do_parallel = TRUE,
  cores = NA,
  verbose = T
)
```

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#### **Arguments**

gobject giotto object

bin\_method method to binarize gene expression

expression\_values

expression values to use

subset\_genes only select a subset of genes to test

spatial\_network\_name

name of spatial network to use (default = 'spatial\_network')

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

percentage\_rank

percentage of top cells for binarization

do\_fisher\_test perform fisher test

get\_av\_expr calculate the average expression per gene of the high expressing cells

get\_high\_expr calculate the number of high expressing cells per gene

do\_parallel run calculations in parallel with mclapply cores number of cores to use if do\_parallel = TRUE

verbose be verbose

#### **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 (optional):

- Number of cells with high expression (binary = 1)
- total of high expressing cells

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

#### Value

data.table with results (see details)

# **Examples**

binGetSpatialGenes(gobject)

 $bin {\tt GetSpatialGenesOld} \ \ \textit{bin GetSpatialGenesOld}$ 

# **Description**

Rapid computation of genes that are spatially clustered

### Usage

```
binGetSpatialGenesOld(
  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
)
```

#### **Arguments**

```
giotto object
gobject
                  method to binarize gene expression
bin_method
expression_values
                  expression values to use
                  only select a subset of genes to test
subset_genes
spatial_network_name
                  name of spatial network to use (default = 'spatial_network')
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

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• 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**

```
binGetSpatialGenesOld(gobject)
```

calculateHVG

calculateHVG

# **Description**

compute highly variable genes

```
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
)
```

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### **Arguments**

gobject giotto object

expression\_values

expression values to use

method method to calculate highly variable genes

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\_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

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

## **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)

24 calculateMetaTableCells

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

calculateMetaTableCells

# Description

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

```
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)
```

### **Description**

Calculate spatial genes using distance matrix.

# Usage

```
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.

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#### Value

data.table with spatial scores

# **Examples**

```
calculate_spatial_genes_python(gobject)
```

```
cellProximityBarplot cellProximityBarplot
```

# **Description**

Create barplot from cell-cell proximity scores

# Usage

```
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
min_orig_ints
                  filter on minimum original cell-cell interactions
min_sim_ints
                  filter on minimum simulated cell-cell interactions
p_val
                  p-value
                  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**

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

#### Value

ggplot barplot

#### **Examples**

```
cellProximityBarplot(CPscore)
```

```
cellProximityEnrichment
```

cell Proximity Enrichment

# **Description**

Compute cell-cell interaction enrichment (observed vs expected)

#### Usage

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

# **Arguments**

### **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")
```

# Arguments

```
giotto object
gobject
                  CPscore, output from cellProximityEnrichment()
CPscore
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
                  character color vector of length 3
color_names
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 heatmap that shows the spatial proximity enrichment or depletion of cell type pairs.

# Value

```
ggplot heatmap
```

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#### **Examples**

```
cellProximityHeatmap(CPscore)
```

```
cell Proximity Network \qquad cell Proximity Network
```

# **Description**

Create network from cell-cell proximity scores

### Usage

```
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 = c("Fruchterman", "DrL", "Kamada-Kawai"),
  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 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
node_size
                 size of nodes
node_text_size size of node labels
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
```

# **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

# **Examples**

```
cellProximityNetwork(CPscore)
```

```
cellProximitySpatPlot cellProximitySpatPlot
```

# **Description**

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

# Usage

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

# **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
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
point_other_border_col
                  border color of other points
point_other_border_stroke
                  stroke size of other points
                  show plots
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

#### See Also

```
cellProximitySpatPlot2D and cellProximitySpatPlot3D for 3D
```

#### **Examples**

```
cellProximitySpatPlot(gobject)
```

```
cellProximitySpatPlot2D
```

cellProximitySpatPlot2D

# Description

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

# Usage

```
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"
)
```

#### **Arguments**

```
cluster_column cluster column with cell clusters
                  x-axis dimension name (default = 'sdimx')
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
                  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
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 SpatPlot 3D \\ cell Proximity SpatPlot 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
                  x-axis dimension name (default = 'sdimx')
sdimx
sdimy
                  y-axis dimension name (default = '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_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
show_legend
                  show legend
point_size_select
                  size of selected points
point_size_other
                  size of other points
                  show plots
show_plot
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)
```

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```
cellProximityVisPlot cellProximityVisPlot
```

# Description

Visualize cell-cell interactions according to spatial coordinates

# Usage

```
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",
  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

cellProximityVisPlot 37

sdimx x-axis dimension name (default = 'sdimx')
sdimy y-axis dimension name (default = '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\_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

point\_other\_border\_col

border color of other points

point\_other\_border\_stroke

stroke size of other 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

#### Usage

```
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

```
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

point\_other\_border\_col

border color of other points

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

### Usage

```
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**

```
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
point_size_select
                  size of selected points
coord_fix_ratio
                  fix ratio between x and y-axis
```

### **Details**

Description of parameters.

### Value

plotly

#### **Examples**

```
cell Proximity VisPlot\_2D\_plotly (gobject)
```

### **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**

```
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
                  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
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 changeGiottoInstructions}
```

change Giot to Instructions

## Description

Function to change one or more instructions from giotto object

# Usage

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

# 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()

44 clusterCells

clusterCells

clusterCells

#### **Description**

cluster cells using a variety of different methods

#### Usage

```
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,
  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,
```

clusterCells 45

```
seed_number = 1234
)
```

### **Arguments**

gobject giotto object

cluster\_method community cluster method to use

name 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\_omega louvain param: omega

walk\_steps randomwalk: number of steps
walk\_clusters randomwalk: number of clusters
walk\_weights randomwalk: weight column

sNNclust\_k SNNclust: k neighbors to use

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

km\_algorithm kmeans algorithm

 $hc\_agglomeration\_method$ 

hierarchical clustering method

hc\_k hierachical number of clusters

hc\_h hierarchical tree cutoff

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

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

## **Examples**

```
clusterCells(gobject)
```

clusterSpatialCorGenes

*clusterSpatialCorGenes* 

### **Description**

Cluster based on spatially correlated genes

## Usage

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

combineMetadata 47

### **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

### Value

Extended cell metadata in data.table format.

## **Examples**

combineMetadata(gobject)

48 combine\_ints\_f

```
combine_ints_f
combine_ints_f
```

# Description

function to combine gene enrichment interactions

## Usage

```
combine_ints_f(
  cell_int,
  all_ints,
  unif_gene_scores,
  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
)
```

# Arguments

```
cell_int
                  selected cell interaction
all_ints
                  all interactions
unif_gene_scores
                  unif_gene_scores results
specific_genes_1
                  specific source genes (see details)
specific_genes_2
                  specific target genes (see details)
min_cells
                  min number of cells threshold
min_spat_diff
                  spatial difference threshold
min_log2_fc
                  log2 fold-change threshold
min_pval
                  p-value threshold
```

## Value

Gene to gene scores in data.table format

```
convertEnsemblToGeneSymbol
```

convertEnsemblToGeneSymbol

### **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)
```

```
createGiottoInstructions
```

create Giot to Instructions

### **Description**

Function to set global instructions for giotto functions

### 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
```

50 createGiottoObject

### **Arguments**

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

#### Value

named vector with giotto instructions

### **Examples**

createGiottoInstructions()

# Description

Function to create a giotto object

# Usage

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

createGiottoObject 51

#### **Arguments**

matrix with raw expression counts [required] raw\_exprs 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 list of instructions or output result from createGiottoInstructions instructions

## 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.

[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

52 createHeatmap\_DT

#### Value

```
giotto object
```

### **Examples**

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

createHeatmap\_DT

createHeatmap\_DT

### **Description**

creates order for clusters

## Usage

```
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"
)
```

### **Arguments**

```
gobject
                 giotto object
expression_values
                 expression values to use
genes
                 genes to use
cluster_column name of column to use for clusters
                 method to determine cluster order
cluster_order
cluster_custom_order
                 custom order for clusters
cluster_cor_method
                 method for cluster correlation
cluster_hclust_method
                 method for hierarchical clustering of clusters
                 method to determine gene order
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
```

createMetagenes 53

#### **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.

#### Usage

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

### **Arguments**

```
gobject Giotto object
expression_values
expression values to use
gene_clusters numerical vector with genes as names
name name of the metagene results
return_gobject return giotto object
```

### 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

```
createMetagenes(gobject)
```

54 createNearestNetwork

createNearestNetwork createNearestNetwork

## Description

create a nearest neighbour (NN) network

## Usage

```
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,
  ...
)
```

### **Arguments**

```
giotto object
gobject
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
                 if dim_reduction_to_use = NULL, which genes to use
genes_to_use
expression_values
                 expression values to use
                 arbitrary name for NN network
name
return_gobject boolean: return giotto object (default = TRUE)
                 number of k neighbors to use
minimum_shared minimum shared neighbors
top_shared
                 keep at ...
                 be verbose
verbose
                 additional parameters for kNN and sNN functions from dbscan
. . .
```

createSpatialEnrich 55

#### **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

#### 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.

56 createSpatialEnrich

#### Usage

```
createSpatialEnrich(
  gobject,
  enrich_method = c("PAGE", "rank", "hypergeometric"),
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  p_value = TRUE,
  n_genes = 100,
  n_times = 1000,
  top_percentage = 5,
  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
p_value
                  calculate p-value (default = TRUE)
n_genes
                  (page/rank) number of randomly selected genes for each permuation
                  (page/rank) number of permutation iterations to calculate p-value
n_times
top_percentage (hyper) percentage of cells that will be considered to have gene expression with
                  matrix binarization
output_enrichment
                  how to return enrichment output
name
                  to give to spatial enrichment results, default = PAGE
return_gobject return giotto object
```

#### **Details**

For details see the individual functions:

PAGE: PAGEEnrichPAGE: rankEnrichPAGE: hyperGeometricEnrich

### Value

Giotto object or enrichment results if return\_gobject = FALSE

createSpatialGrid 57

#### **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

```
createSpatialGrid(gobject)
```

```
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 59

```
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)
```

60 createSpatialNetwork

```
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 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)
```

```
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**

# 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)
```

create\_cluster\_matrix 63

```
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

64 decide\_cluster\_order

```
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**

```
gobject
                  giotto object
method
                  method to use for spatial averaging
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**

For method = network, it expects a fully connected spatial network. You can make sure to create a fully connected network by setting minimal\_k > 0 in the createSpatialNetwork function.

- 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()

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#### 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
```

dimCellPlot 67

#### **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

dim Cell Plot

#### **Description**

Visualize cells according to dimension reduction coordinates

### Usage

```
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 = NULL,
  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",
```

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```
edge_alpha = NULL,
      point_shape = c("border", "no_border"),
      point_size = 1,
      point_border_col = "black",
      point_border_stroke = 0.1,
      show_legend = T,
      legend_text = 8,
      legend_symbol_size = 1,
      background_color = "white",
      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
                    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
    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 subset of cells based on cell IDs
    select_cells
    show_other_cells
                     display not selected cells
```

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```
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_shape
                  point with border or not (border or no_border)
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
axis_text
                  size of axis title
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
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
                  title for plot, defaults to cell_color parameter
title
```

### **Details**

Description of parameters. For 3D plots see dimCellPlot2D

## Value

ggplot

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#### **Examples**

```
dimCellPlot(gobject)
```

dimCellPlot2D

dim Cell Plot 2D

#### **Description**

Visualize cells according to dimension reduction coordinates

#### Usage

```
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 = NULL,
  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_shape = c("border", "no_border"),
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  background_color = "white",
  axis_text = 8,
  axis_title = 8,
  cow_n_col = 2,
  cow_rel_h = 1,
```

dimCellPlot2D 71

```
cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimCellPlot2D"
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 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
```

size of labels

label\_size

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```
label_fontface font of labels
                  column to use for alpha of the edges
edge_alpha
point_shape
                  point with border or not (border or no_border)
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
axis_text
                  size of axis title
axis_title
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
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

```
dimCellPlot2D(gobject)
```

dimGenePlot 73

dimGenePlot

dimGenePlot

### **Description**

Visualize cells and gene expression according to dimension reduction coordinates

# Usage

```
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,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  scale_alpha_with_expression = FALSE,
  point_shape = c("border", "no_border"),
  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,
  show_legend = T,
  legend_text = 8,
  background_color = "white",
  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 = "dimGenePlot"
)
```

# Arguments

gobject giotto object

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```
expression_values
                 gene expression values to use
                 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)
show_legend
                 show legend
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_plot
                 show plots
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
                 parameters for cowplot::save_plot()
```

# **Details**

Description of parameters.

### Value

ggplot

### See Also

dimGenePlot3D

dimGenePlot2D 75

### **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",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  scale_alpha_with_expression = FALSE,
  point_shape = c("border", "no_border"),
  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,
  show_legend = T,
  legend_text = 8,
  background_color = "white",
  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 = "dimGenePlot2D"
```

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# 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 with border or not (border or no border) point\_shape size of point (cell) point\_size point\_border\_col color of border around points point\_border\_stroke stroke size of border around points midpoint size of point (cell) show legend show\_legend legend\_text size of legend text background\_color color of plot background axis\_text size of axis text size of axis title 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 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() . . .

dimGenePlot3D 77

### **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",
  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,
```

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```
save_plot = NA,
save_param = list(),
default_save_name = "dimGenePlot3D"
)
```

# **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
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
                 column to use for alpha of the edges
edge_alpha
point_size
                 size of point (cell)
show_legend
                 show legend
                 show plots
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
                 parameters for cowplot::save_plot()
```

### **Details**

Description of parameters.

# Value

ggplot

### **Examples**

```
dimGenePlot3D(gobject)
```

dimPlot 79

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,
  point_shape = c("border", "no_border"),
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  background_color = "white",
  axis_text = 8,
  axis_title = 8,
  title = NULL,
```

80 dimPlot

```
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
    {\tt 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)
                     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
    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

dimPlot 81

```
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_shape
                  point with border or not (border or no_border)
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
axis_text
                  size of axis text
axis_title
                  size of axis title
title
                  title for plot, defaults to cell_color parameter
                  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_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
groub_by
                  create multiple plots based on cell annotation column
```

#### **Details**

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

# Value

ggplot

# **Examples**

```
dimPlot(gobject)
```

82 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_shape = c("border", "no_border"),
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  title = NULL,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  background_color = "white",
  axis_text = 8,
  axis_title = 8,
```

dimPlot2D 83

```
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 = "dimPlot2D"
    )
Arguments
    gobject
                     giotto object
    group_by_subset
                     subset the group_by factor column
    {\tt 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)
                     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
    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

84 dimPlot2D

```
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 with border or not (border or no_border)
point_shape
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
axis_text
axis_title
                  size of axis title
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
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 85

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_shape = c("border", "no_border"),
  point_size = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  title = NULL,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  background_color = "white",
  axis_text = 8,
  axis_title = 8,
  show_plot = NA,
  return_plot = NA,
```

86 dimPlot2D\_single

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

dimPlot3D 87

```
column to use for alpha of the edges
edge_alpha
point_shape
                  point with border or not (border or no_border)
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
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 for saving, don't change, change save\_name in save\_param

### **Details**

Description of parameters. For 3D plots see dimPlot3D

# Value

ggplot

default\_save\_name

# **Examples**

dimPlot2D\_single(gobject)

dimPlot3D dimPlot3D

# **Description**

Visualize cells according to dimension reduction coordinates

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### Usage

```
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,
      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
    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
   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

dimPlot3D

```
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
                 size of point (cell)
point_size
                 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
show_legend
                 show legend
```

### **Details**

Description of parameters.

### Value

plotly

# **Examples**

dimPlot3D(gobject)

90 doHclust

```
direction_test_CPG direction_test_CPG
```

# **Description**

shows direction of change

### Usage

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

### **Examples**

```
direction_test_CPG()
```

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
)
```

### **Arguments**

```
gobject giotto object
expression_values
expression values to use
genes_to_use subset of genes to use
```

doHMRF 91

```
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimensions reduction name
dimensions_to_use
                 dimensions to use
distance_method
                 distance method
agglomeration_method
                 agglomeration method for hclust
k
                 number of final clusters
                 cut hierarchical tree at height = h
h
                 name for hierarchical clustering
name
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

hclust

# **Examples**

doHclust(gobject)

doHMRF

doHMRF

# Description

Run HMRF

```
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",
```

92 doHMRF

```
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
)
```

# Arguments

```
giotto object
gobject
expression_values
                 expression values to use
spatial_network_name
                 name of spatial network to use for HMRF
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 of HMRF run
name
                 number of HMRF domains
k
betas
                 betas to test for
tolerance
                 tolerance
zscore
                 zscore
numinit
                 number of initializations
                 python path to use
python_path
output_folder
                 output folder to save results
overwrite_output
```

#### **Details**

Description of HMRF parameters ...

## Value

Creates a directory with results that can be viewed with viewHMRFresults

overwrite output folder

# **Examples**

```
doHMRF(gobject)
```

doKmeans 93

doKmeans doKmeans

# **Description**

cluster cells using kmeans algorithm

# Usage

```
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",
    "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
                 subset of genes to use
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
                 number of final clusters
centers
                 kmeans maximum iterations
iter_max
nstart
                 kmeans nstart
                 kmeans algorithm
algorithm
name
                 name for kmeans clustering
return_gobject boolean: return giotto object (default = TRUE)
set_seed
                 set seed
seed_number
                 number for seed
```

94 doLeidenCluster

### **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)
```

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,
  ...
)
```

# **Arguments**

doLeidenSubCluster 95

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

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

#### **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 dol

doLeidenSubCluster

# **Description**

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

96 doLeidenSubCluster

#### **Usage**

nn\_network\_to\_use

network\_name

verbose

```
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",
      return_gobject = TRUE,
      verbose = T
Arguments
    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
                     minimum number of HVG, or all genes will be used as input for PCA
   min_nr_of_hvg
                     parameters for runPCA
    pca_param
   nn_param
                     parameters for parameters for createNearestNetwork
                     number of k for createNearestNetwork
   k_neighbors
                     resolution of Leiden clustering
    resolution
    n_iterations
                     number of interations to run the Leiden algorithm.
   python_path
                     specify specific path to python if required
```

type of NN network to use (kNN vs sNN)

name of NN network to use

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

verbose

doLouvainCluster 97

### **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

### See Also

```
doLeidenCluster
```

# **Examples**

```
doLeidenSubCluster(gobject)
```

doLouvainCluster

doLouvainCluster

# **Description**

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

```
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,
)
```

### **Arguments**

gobject giotto object

version implemented version of Louvain clustering to use

name name for cluster

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

python\_path [community] specify specific path to python if required

resolution [community] resolution

gamma [multinet] Resolution parameter for modularity in the generalized louvain method.

omega [multinet] Inter-layer weight parameter in the generalized louvain method.

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)

doLouvainCluster\_community

 $doLouvainCluster\_community$ 

# 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,
   ...
)
```

### **Arguments**

giotto object gobject 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 specify specific path to python if required resolution resolution weight\_col weight column to use for edges Will randomize the node evaluation order and the community evaluation order louv\_random to get different partitions at each call return\_gobject boolean: return giotto object (default = TRUE) set\_seed set seed

### **Details**

seed\_number

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

number for seed

### **Examples**

```
doLouvainCluster_community(gobject)
```

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

# Description

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

# Usage

```
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,
  ...
)
```

# Arguments

```
gobject
                  giotto object
                  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
                  Resolution parameter for modularity in the generalized louvain method.
gamma
                  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**

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 101

doLouvainSubCluster doLouvainSubCluster

# **Description**

subcluster cells using a NN-network and the Louvain algorithm

### Usage

```
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,
  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 of Louvain algorithm to use
version
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
```

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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 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{lower} 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
)
```

# **Arguments**

```
gobject
                  giotto object
                  name for new clustering result
name
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
```

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)

doLouvainSubCluster\_multinet

doLouvainSubCluster\_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
Arguments
                   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

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

number of k for createNearestNetwork k\_neighbors

gamma gamma omega omega 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

specify specific path to python if required python\_path

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### **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)
```

doRandomWalkCluster

doRandomWalkCluster

# Description

Cluster cells using a random walk approach.

```
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,
  ...
)
```

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# **Arguments**

```
giotto object
gobject
                 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
walk_steps
                 number of walking steps
walk_clusters
                 number of final clusters
walk_weights
                 cluster column defining the 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

doSNNCluster

# Description

Cluster cells using a SNN cluster approach.

```
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)
```

```
do_spatial_grid_averaging
```

 $do\_spatial\_grid\_averaging$ 

# Description

smooth gene expression over a defined spatial grid

```
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
)
```

```
gobject giotto object

expression_values

gene expression values to use

subset_genes subset of genes to use

spatial_network_name

name of spatial network to use

b smoothing factor beteen 0 and 1 (default: automatic)
```

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#### **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)
```

DT\_removeNA

DT\_removeNA

# Description

set NA values to 0

## Usage

DT\_removeNA(DT)

 $dt_{to_matrix}$ 

 $dt\_to\_matrix$ 

# **Description**

converts data.table to matrix

## Usage

```
dt_to_matrix(x)
```

# Examples

dt\_to\_matrix(x)

exportGiottoViewer 111

exportGiottoViewer

*exportGiottoViewer* 

#### **Description**

compute highly variable genes

#### Usage

```
exportGiottoViewer(
  gobject,
  output_directory = NULL,
  spat_enr_names = NULL,
  factor_annotations = NULL,
  numeric_annotations = NULL,
  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
)
```

```
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

## 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.

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#### Value

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

## **Examples**

```
{\tt 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

```
extractNearestNetwork 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

114 filterCombinations

## **Examples**

```
extractNearestNetwork(gobject)
```

fDataDT

fDataDT

## **Description**

show gene metadata

## Usage

```
fDataDT(gobject)
```

## **Arguments**

gobject

giotto object

#### Value

data.table with gene metadata

## **Examples**

```
pDataDT(gobject)
```

filterCombinations

filter Combinations

## **Description**

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

```
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
)
```

filterCPGscores 115

#### **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-
min_det_genes_per_cell
                  minimum number of expressed genes per cell to consider that cell further
                  ggplot transformation for x-axis (e.g. log2)
scale_x_axis
x_axis_offset x-axis offset to be used together with the scaling transformation
scale_y_axis
                  ggplot transformation for y-axis (e.g. log2)
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.

#### Value

list of data.table and ggplot object

## **Examples**

```
filterCombinations(gobject)
```

 ${\tt filter CPG scores}$ 

filterCPGscores

#### **Description**

visualize Cell Proximity Gene enrichment scores

```
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")
)
```

116 filterDistributions

#### **Arguments**

#### **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

## **Examples**

```
filterCPGscores(CPGscore)
```

```
filterDistributions filterDistributions
```

## **Description**

show gene or cell distribution after filtering on expression threshold

```
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
)
```

filterGiotto 117

## **Arguments**

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

#### Value

ggplot object

# **Examples**

filterDistributions(gobject)

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
)
```

```
gobject giotto object
expression_values
expression values to use
expression_threshold
threshold to consider a gene expressed
```

118 findGiniMarkers

#### **Details**

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

## Value

giotto object

## **Examples**

```
filterGiotto(gobject)
```

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 = 0.2,
  min_det_gini_score = 0.2,
  detection_threshold = 0,
  rank_score = 1,
  min_genes = 5
)
```

```
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
```

#### **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 score = detection rank x expression rank x expr gini-coefficient x detection gini-coefficient
- 7. for each gene sort on expression and detection rank and combined score

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,
  rank_score = 1,
  min_genes = 4,
  verbose = TRUE
)
```

## **Arguments**

```
giotto object
gobject
expression_values
                  gene expression values to use
cluster_column clusters to use
subset_clusters
                  selection of clusters to compare
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
rank_score
                 rank scores for both detection and expression to include
min_genes
                  minimum number of top genes to return
verbose
                  be verbose
```

### Value

data.table with marker genes

## See Also

findGiniMarkers

```
findGiniMarkers_one_vs_all(gobject)
```

findMarkers 121

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,
 min\_genes = 4,
 group_1_name = NULL,
  group_2_name = NULL,
 adjust_columns = NULL,
)
```

findMarkers

```
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
{\tt detection\_threshold}
                  gini: detection threshold for gene expression
                  gini: rank scores to include
rank_score
                  minimum number of top genes to return (for gini)
min_genes
                  mast: custom name for group_1 clusters
group_1_name
```

```
group_2_name mast: custom name for group_2 clusters

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

findScranMarkers, findGiniMarkers and findMastMarkers

#### **Examples**

```
findMarkers(gobject)
```

# Description

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

```
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,
)
```

findMarkers\_one\_vs\_all 123

#### **Arguments**

gobject giotto object

expression\_values

gene expression values to use

cluster\_column clusters to use

subset\_clusters

selection of clusters to compare

method method to use to detect differentially expressed genes

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

rank\_score gini: rank scores to include

adjust\_columns mast: column in pDataDT to adjust for (e.g. detection rate)

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

findScranMarkers\_one\_vs\_all, findGiniMarkers\_one\_vs\_all and findMastMarkers\_one\_vs\_all

```
findMarkers_one_vs_all(gobject)
```

124 findMastMarkers

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,
  ...
)
```

## **Arguments**

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
                  group 1 cluster IDs from cluster_column for pairwise comparison
group_1
                  custom name for group_1 clusters
group_1_name
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
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

```
findMastMarkers(gobject)
```

```
find {\it MastMarkers\_one\_vs\_all} \\ find {\it MastMarkers\_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,
  ...
)
```

## **Arguments**

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

## Value

data.table with marker genes

## See Also

findMastMarkers

```
findMastMarkers_one_vs_all(gobject)
```

126 findScranMarkers

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,
  ...
)
```

# 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 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

```
findScranMarkers(gobject)
```

```
findScranMarkers_one_vs_all findScranMarkers_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,
  ...
)
```

## **Arguments**

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

## Value

data.table with marker genes

## See Also

findScranMarkers

```
findScranMarkers_one_vs_all(gobject)
```

128 find\_grid\_y

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

 $find\_grid\_y$ 

# Description

find grid location on y-axis

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

find\_grid\_z

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

```
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 131

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)
```

132 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 133

get10Xmatrix

get10Xmatrix

# Description

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

# Usage

```
get10Xmatrix(path_to_data, row_ids = c("gene_symbols", "ensembl_ids"))
```

# Arguments

```
path_to_data path to the 10X folder
row_ids row ids to use (gene symbols or ensembl ids)
```

#### **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)
```

```
{\it getCellProximityGeneScores} \\ {\it getCellProximityGeneScores}
```

# Description

Compute cell-cell interaction enrichment (observed vs expected)

#### Usage

verbose

verbose

```
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"),
      false_discovery_test = c("holm", "hochberg", "hommel", "bonferroni", "BH", "BY",
        "fdr", "none"),
      false_discovery_target = c("cell_interactions", "genes"),
      minimum_unique_cells = NA,
      fold_change_addendum = 0.1,
      in_two_directions = TRUE,
      exclude_selected_cells_from_test = F,
      do_parallel = TRUE,
      cores = NA,
      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
                    perform differential test
    do_diff_test
    diff_test
                     which differential expression test
    false_discovery_test
                     test to adjust p-values for multiple hypothesis testing
    false_discovery_target
                     adjust p-values per cell-cell pair or per gene
   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
    do_parallel
                     run enrichment calculations in parallel with mclapply
                     number of cores to use if do_parallel = TRUE
    cores
```

#### **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
- 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)

136 getDendrogramSplits

```
getClusterSimilarity
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")
)
```

## **Arguments**

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

#### **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..

getDistinctColors 137

#### 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
)
```

## **Arguments**

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

#### **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
```

#### **Description**

Returns a number of distint colors based on the RGB scale

```
getDistinctColors(n)
```

#### **Arguments**

n number of colors wanted

#### Value

number of distinct colors

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,
   do_parallel = TRUE,
   cores = NA,
   verbose = TRUE
)
```

```
CPGscore, output from getCellProximityGeneScores()
CPGscore
selected_genes select subset of genes
specific_genes_1
                  specific source genes (see details)
specific_genes_2
                  specific target genes (see details)
min_cells
                  min number of cells threshold
min_spat_diff
                  spatial difference threshold
min_log2_fc
                  log2 fold-change threshold
direction
                  up or downregulation or both
fold_change_addendum
                  constant to add when calculating log2 fold-change
do_parallel
                  run enrichment calculations in parallel with mclapply
cores
                  number of cores to use if do_parallel = TRUE
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)
```

## **Examples**

```
{\tt get\_cell\_to\_cell\_sorted\_name\_conversion()}
```

## **Description**

Computes gene enrichment between all interactions

```
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,
do_parallel = TRUE,
cores = NA,
verbose = T
```

# **Examples**

```
get_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,
   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
)
```

```
get_specific_interaction_gene_enrichment()
```

ggplot\_save\_function 141

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

142 giotto-class

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

 $heatmSpatialCorGenes \quad \textit{heatmSpatialCorGenes}$ 

#### **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
use_clus_name
                 name of clusters to visualize (from clusterSpatialCorGenes())
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
                 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
                 additional parameters to the Heatmap function from ComplexHeatmap
. . .
```

#### 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,
  top_percentage = 5,
  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

top_percentage percentage of cells that will be considered to have gene expression with matrix binarization

output_enrichment

how to return enrichment output
```

#### **Details**

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

#### Value

data.table with enrichment results

kmeans\_binarize 145

# **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

```
\begin{array}{ccc} 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 \end{array}
```

betas\_used betas that were tested

 $\verb"python_path_used"$ 

python path that was used

## **Details**

Description of HMRF parameters ...

## Value

reloads a previous ran HMRF from doHRMF

```
loadHMRF(gobject)
```

146 makeSignMatrixRank

makeSignMatrixPAGE makeSignMatrixPAGE

### **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. Each cell type or process should have a vector of cell-type or process specific genes. These vectors need to be combined into a list (sign\_list). The names of the cell types or processes that are provided in the list need to be given (sign\_names).

## 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()

 $make Sign Matrix Rank \qquad make Sign Matrix Rank$ 

# 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)
```

#### **Arguments**

sc\_matrix matrix of single-cell RNAseq expression data

sc\_cluster\_ids vector of cluster ids

gobject if giotto object is given then only genes present in both datasets will be consid-

ered

### 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.

## 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,
```

148 mygini\_fun

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

### Arguments

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

### **Examples**

```
mergeClusters(gobject)
```

mygini\_fun

mygini\_fun

# Description

calculate gini coefficient

## Usage

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

# Value

gini coefficient

nnDT\_to\_kNN 149

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

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

# Value

list of splitted dendrogram nodes from high to low node height

```
node_clusters(hclus_obj)
```

normalizeGiotto

normalizeGiotto

normalizeGiotto

## **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
)
```

## **Arguments**

```
giotto object
gobject
norm_methods
                  normalization method to use
library_size_norm
                  normalize cells by library size
scalefactor
                  scale factor to use after library size normalization
                  transform values to log-scale
log_norm
                  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
```

### **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:

normalizeGiottoFast 151

• 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)
```

normalize Giotto Fast

normalizeGiottoFast

## **Description**

fast normalize and/or scale expresion values of Giotto object

### Usage

```
normalizeGiottoFast(
  gobject,
  norm_methods = c("standard", "osmFISH"),
  library_size_norm = TRUE,
  scalefactor = 6000,
  log_norm = TRUE,
  log_offset = 1,
  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
scalefactor
                  scale factor to use after library size normalization
log_norm
                  transform values to log-scale
log_offset
                  offset value to add to expression matrix, default = 1
logbase
                  log base to use to log normalize expression values
                  z-score genes over all cells
scale_genes
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**

normalizeGiottoFast(gobject)

OR\_function2

OR\_function2

# Description

calculate odds-ratio

#### Usage

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

PAGEEnrich 153

**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**

### **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)
```

pDataDT

pagePermutation

pagePermutation

# Description

creates permutation for the PAGEEnrich test

# Usage

```
pagePermutation(sc_gene, gene_number, n)
```

# Examples

pagePermutation()

pDataDT

pDataDT

# Description

show cell metadata

# Usage

pDataDT(gobject)

# Arguments

gobject

giotto object

## Value

data.table with cell metadata

# **Examples**

pDataDT(gobject)

plotCPGscores 155

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
)
```

# **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
                  show a simplified plot
simple_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
show_plot
                  show plot
```

### **Details**

Give more details ...

### Value

ggplot barplot

```
plotCPGscores(CPGscores)
```

156 plotGTGscores

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

```
giotto object
gobject
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
                  ggplot facet scales paramter
facet_scales
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]
                  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
selected_genes genes to show
```

plotHeatmap 157

### **Details**

Give more details ...

#### Value

ggplot barplot

## **Examples**

plotGTGscores(GTGscore)

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"
```

158 plotHeatmap

### **Arguments**

gobject giotto object expression\_values expression values to use genes to use genes 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 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

plotly\_axis\_scale\_2D 159

### Value

ggplot

# **Examples**

```
plotHeatmap(gobject)
```

```
{\tt plotly\_axis\_scale\_2D} \quad \textit{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()
```

```
plotly_axis_scale_2D(gobject)
```

plotly\_grid

```
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
)
```

### **Arguments**

```
cell_locations spatial_loc in giotto object sdimx x axis of cell spatial location sdimy y axis of cell spatial location sdimz z 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_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"
)
```

plotly\_network 161

# **Arguments**

```
spatial_grid spatial_grid in giotto object
```

### Value

```
edges in spatial grid as data.table()
```

## **Examples**

```
plotly_grid(gobject)
```

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()
```

```
plotly_network(gobject)
```

```
plot {\tt MetaDataCellsHeatmap} \\ plot {\tt MetaDataCellsHeatmap}
```

## 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,
  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"
)
```

```
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
                  custom gene order (default = NULL)
gene_cor_method
                  correlation method for genes
gene_cluster_method
                  hierarchical cluster method for the genes
```

### **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.

```
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"
)
```

```
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

show_values which values to show on heatmap

custom_cluster_order

custom cluster order (default = NULL)
```

plotMetaDataHeatmap 165

```
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
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 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
```

## **Details**

Creates heatmap for the average expression of selected genes in the different annotation/cluster groups. Calculation of cluster or gene order is done on the provided expression values, but visualization is by default on the z-scores. Other options are the original values or z-scores rescaled per gene (-1 to 1).

## Value

ggplot or data.table

# See Also

plotMetaDataCellsHeatmap for numeric cell annotation instead of gene expression.

# **Examples**

plotMetaDataHeatmap(gobject)

166 plotPCA

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
                  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
```

plotPCA 167

```
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_shape
                  point with border or not (border or no_border)
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
axis_text
                  size of axis title
axis_title
                  cowplot param: how many columns
cow_n_col
                  cowplot param: relative height
cow_rel_h
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
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
```

## **Details**

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

### Value

ggplot

```
plotPCA(gobject)
```

168 plotPCA\_2D

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
```

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```
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_shape
                  point with border or not (border or no_border)
                  size of point (cell)
point_size
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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
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
                  cowplot param: relative width
cow_rel_w
                  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
```

# Details

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

### Value

ggplot

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#### **Examples**

```
plotPCA_2D(gobject)
```

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
                  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)
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 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
```

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```
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)
                  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", ...)
```

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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 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\_shape point with border or not (border or no\_border) 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

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```
legend_symbol_size
                 size of legend symbols
background_color
                 color of plot background
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
                 cowplot param: relative width
cow_rel_w
                 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
```

# **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",
  ...
)
```

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### **Arguments**

giotto object gobject 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 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

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point with border or not (border or no\_border) point\_shape size of point (cell)

point\_size

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

title for plot, defaults to cell\_color parameter title

show\_legend show legend

legend\_text size of legend text

 ${\tt legend\_symbol\_size}$ 

size of legend symbols

background\_color

color of plot background

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 cowplot param: how to align cow\_align

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

## **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

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```
Usage
```

label\_size

edge\_alpha

point\_size
show\_legend

size of labels

show legend

size of point (cell)

column to use for alpha of the edges

```
plotTSNE_3D(
      gobject,
      dim_reduction_name = "tsne",
      default_save_name = "TSNE_3D",
Arguments
    gobject
                      giotto object
    dim_reduction_name
                      tsne 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
```

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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**

Description of parameters.

### Value

plotly

## **Examples**

```
plotTSNE_3D(gobject)
```

plotUMAP plotUMAP

# Description

Short wrapper for UMAP visualization

### Usage

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

### **Arguments**

gobject giotto object

dim\_reduction\_name

dimension reduction name

 ${\tt default\_save\_name}$ 

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 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)

cell\_color color for cells (see details)

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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\_shape point with border or not (border or no\_border) size of point (cell) point\_size 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 legend\_symbol\_size size of legend symbols background\_color color of plot background 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

cowplot param: relative width

cow\_rel\_w

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```
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

### **Details**

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

### Value

ggplot

### **Examples**

```
plotUMAP(gobject)
```

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
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
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
```

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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  ${\tt cell\_color\_gradient}$ 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 size of labels label\_size label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_shape point with border or not (border or no\_border) point\_size size of point (cell) point\_border\_col color of border around points point\_border\_stroke stroke size of border around points title for plot, defaults to cell\_color parameter title show\_legend show legend legend\_text size of legend text legend\_symbol\_size size of legend symbols background\_color color of plot background size of axis text axis\_text

plotUMAP\_3D

```
size of axis title
axis_title
                 cowplot param: how many columns
cow_n_col
cow_rel_h
                 cowplot param: relative height
                 cowplot param: relative width
cow_rel_w
                 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
```

#### **Details**

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

#### Value

ggplot

### **Examples**

```
plotUMAP_2D(gobject)
```

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",
   ...
)
```

## Arguments

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```
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
label_size
                  size of labels
                  column to use for alpha of the edges
edge_alpha
                  size of point (cell)
point_size
show_legend
                  show legend
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
```

### **Details**

Description of parameters.

#### Value

plotly

## Examples

```
plotUMAP_3D(gobject)
```

```
plot\_network\_layer\_ggplot \\ plot\_network\_layer\_ggplot
```

## 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
                     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
{\tt 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)
```

```
plot\_point\_layer\_ggplot\_noFILL \\ plot\_point\_layer\_ggplot\_noFILL
```

## Description

Visualize cells in point layer according to dimension reduction coordinates without borders

#### Usage

```
plot_point_layer_ggplot_noFILL(
     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,
      show_cluster_center = F,
      show_center_label = T,
      center_point_size = 4,
     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_cells
                    select subset of cells based on cell IDs
                    size of point (cell)
   point_size
   show_cluster_center
```

plot center of selected clusters

plot label of selected clusters

show\_center\_label

```
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
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_noFILL(gobject)
```

### **Description**

creat ggplot point layer for spatial coordinates

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

```
label_size size of labels

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

gobject giotto object
```

### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
plot_spat_point_layer_ggplot(gobject)
```

#### **Description**

creat ggplot point layer for spatial coordinates without borders

```
plot_spat_point_layer_ggplot_noFILL(
  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,
  show_cluster_center = F,
  show_center_label = T,
```

gobject

giotto object

```
center_point_size = 4,
      label_size = 4,
      label_fontface = "bold",
      show_other_cells = T,
      other_cell_color = "lightgrey",
      other_point_size = 1,
      show_legend = TRUE
    )
Arguments
                      x-axis dimension name (default = 'sdimx')
    sdimx
                      y-axis dimension name (default = 'sdimy')
    sdimy
    cell_locations_metadata_selected
                      annotated location from selected cells
    {\tt 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 subset of cells based on cell IDs
    select_cells
                      size of point (cell)
    point_size
    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_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
```

print.giotto 191

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

```
plot_spat_point_layer_ggplot_noFILL(gobject)
```

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.

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

192 rankPermutation

### **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 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  $R = (R1*R2)^{(1/2)}$ , 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:  $RBP = (1 - 0.99) * (0.99)^{(R - 1)}$  and the final enrichment score is then calculated as the sum of top 100 RBPs.

#### Value

data.table with enrichment results

#### See Also

 ${\it make Sign Matrix Rank}$ 

#### **Examples**

rankEnrich(gobject)

rankPermutation

rankPermutation

#### **Description**

creates permutation for the rankEnrich test

#### Usage

```
rankPermutation(sc_gene, n)
```

#### **Examples**

rankPermutation()

rankSpatialCorGroups 193

```
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
                  name of clusters to visualize (from clusterSpatialCorGenes())
use_clus_name
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
```

## 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

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

194 removeCellAnnotation

```
{\tt readGiottoInstructions}
```

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 removeCellAnnotation

# 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

removeGeneAnnotation 195

### **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

## 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)

196 runPCA

#### Value

named vector with giotto instructions

### **Examples**

```
replaceGiottoInstructions()
```

runPCA

runPCA

### **Description**

runs a Principal Component Analysis

### Usage

```
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,
  ...
)
```

# 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
ncp
                  number of principal components to calculate
                  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 197

runtSNE runtSNE

#### **Description**

run tSNE

### Usage

```
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,
)
```

# Arguments

```
giotto object
gobject
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
                 arbitrary name for tSNE run
name
                 if dim_reduction_to_use = NULL, which genes to use
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
                 tSNE param: do PCA before tSNE (default = FALSE)
do_PCA_first
                 use of seed
set_seed
seed_number
                 seed number to use
```

additional tSNE parameters

198 runUMAP

#### **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

runUMAP

### **Description**

run UMAP

```
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,
)
```

runUMAP 199

## **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 UMAP run

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

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

n\_neighbors UMAP param: number of neighbors n\_components UMAP param: number of components

n\_epochs UMAP param: number of epochs 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)

200 selectPatternGenes

selectPatternGenes se

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    | Output from detectSpatialPatterns                     |
|---------------|---|
| 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. |

### **Details**

Description.

## Value

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

## **Examples**

```
selectPatternGenes(gobject)
```

# 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

## Usage

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

 $show {\tt ClusterDendrogram} \ \ \textit{show ClusterDendrogram}$ 

## **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
                  height of horizontal lines to plot
h
h_color
                  color of horizontal lines
rotate
                  rotate dendrogram 90 degrees
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.

#### Value

ggplot

## **Examples**

```
showClusterDendrogram(gobject)
```

showClusterHeatmap 203

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**

```
gobject
                  giotto object
expression_values
                  expression values to use
                  vector of genes to use, default to 'all'
genes
cluster_column name of column to use for clusters
cor
                  correlation score to calculate distance
                  distance method to use for hierarchical clustering
distance
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
```

#### **Details**

Correlation heatmap of selected clusters.

#### Value

ggplot

204 showCPGscores

### **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", "heatmap",
    "dotplot"),
 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_cells
                 min number of cells threshold
min_fdr
                 fdr threshold
                 spatial difference threshold
min_spat_diff
min_log2_fc
                 min log2 fold-change
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]
                 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**

Different ways to visualize how many genes are differentially regulated within a source cell type due to the proximity of another neighboring cell type.

### Value

Gene to gene scores in data.table format

### **Examples**

```
showCPGscores(CPGscore)
```

```
show Gene Expression Proximity Score \\ 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

## **Examples**

show Gene Expression Proximity Score (scores)

206 showGTGscores

```
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()
```

 $\verb|showGTGscores||$ 

showGTGscores

### **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 {\it Expression Proximity Score} \\ show Int {\it 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
```

208 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 209

#### 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

gobject

return\_plot

return ggplot object

giotto object

210 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 211

```
grid_border_color
                  color for grid
show_legend
                  show legend of plot
point_size
                  adjust the 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")
```

212 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 213

 $show Spatial Cor Genes \qquad show Spatial Cor Genes$ 

## 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)|
```

214 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 215

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
```

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#### **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 = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
 gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
 point_shape = c("border", "no_border"),
 point_size = 3,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
  center_point_size = 4,
```

spatCellPlot 217

```
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,
legend_symbol_size = 1,
background_color = "white",
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"
```

#### **Arguments**

```
gobject
                  giotto object
                  x-axis dimension name (default = 'sdimx')
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 subset of cells based on cell IDs
select_cells
point_shape
                  point with border or not (border or no_border)
```

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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 spatial grid show\_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 show\_legend show legend legend\_text size of legend text legend\_symbol\_size size of legend symbols background\_color color of plot background axis\_text size of axis text axis\_title size of axis title 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 spatCellPlot2D 219

#### **Details**

Description of parameters.

#### 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 = NULL,
 cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 select_cell_groups = NULL,
  select_cells = NULL,
  point_shape = c("border", "no_border"),
 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",
```

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```
other_point_size = 1,
      other_cells_alpha = 0.1,
      coord_fix_ratio = NULL,
      show_legend = T,
      legend_text = 8,
      legend_symbol_size = 1,
      background_color = "white",
      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')
                     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 subset of cells based on cell IDs
    select_cells
                     point with border or not (border or no_border)
   point_shape
                     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

spatCellPlot2D 221

```
label_size
                  size of labels
label_fontface font of labels
                  show underlying spatial network
show_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
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
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**

```
spatCellPlot2D(gobject)
```

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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 = NULL,
  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_shape = c("border", "no_border"),
  dim_point_size = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border"),
  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",
```

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```
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,
      legend_symbol_size = 1,
      dim_background_color = "white",
      spat_background_color = "white",
      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
    \operatorname{sdim} x
                    = 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

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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\_shape spatial points with border or not (border or no\_border) 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\_shape spatial points with border or not (border or no\_border) 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  $\operatorname{dim\_show\_cluster\_center}$ show the center of each cluster dim\_show\_center\_label provide a label for each cluster  ${\tt dim\_center\_point\_size}$ 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 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 coord\_fix\_ratio ratio for coordinates 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 legend show\_legend legend\_text size of legend text legend\_symbol\_size size of legend symbols dim\_background\_color background color of points in dim. reduction space spat\_background\_color background color of spatial points size of axis text 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 226 spatDimCellPlot2D

#### **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 = NULL,
 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_shape = c("border", "no_border"),
 dim_point_size = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border"),
  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,
```

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```
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,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  dim_background_color = "white",
  spat_background_color = "white",
 axis_text = 8,
 axis_title = 8,
  coord_fix_ratio = 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 = "spatDimCellPlot2D"
)
```

### **Arguments**

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dim\_reduction\_name dimension reduction name dimension to use on x-axis dim1\_to\_use 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 subset of cells based on cell IDs select\_cells dim\_point\_shape dim reduction points with border or not (border or no\_border) 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\_shape spatial points with border or not (border or no\_border) 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 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 show\_legend show legend legend\_text size of legend text legend\_symbol\_size size of legend symbols dim\_background\_color background color of points in dim. reduction space spat\_background\_color background color of spatial points axis\_text size of axis text axis\_title size of axis title coord\_fix\_ratio ratio for coordinates

cowplot param: how many columns

cow\_n\_col

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```
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_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**

```
spatDimCellPlot2D(gobject)
```

spatDimGenePlot
 spatDimGenePlot

### **Description**

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

```
spatDimGenePlot(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 plot_alignment = c("vertical", "horizontal"),
 genes,
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
 dim_point_shape = c("border", "no_border"),
 dim_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,
```

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scale\_alpha\_with\_expression = FALSE,

```
spatial_network_name = "spatial_network",
      spatial_grid_name = "spatial_grid",
      spat_point_shape = c("border", "no_border"),
      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",
      show_legend = T,
      legend_text = 8,
      dim_background_color = "white",
      spat_background_color = "white",
      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 = "spatDimGenePlot"
    )
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
    dim_point_shape
                    dimension points with border or not (border or no_border)
   dim_point_size dim reduction plot: point size
   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
```

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```
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_shape
                  spatial points with border or not (border or no_border)
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)
show_legend
                  show legend
legend_text
                  size of legend text
dim_background_color
                  color of plot background for dimension plot
spat_background_color
                  color of plot background for spatial plot
                  size of axis text
axis_text
axis_title
                  size of 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 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

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#### 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,
 dim_point_shape = c("border", "no_border"),
 dim_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_shape = c("border", "no_border"),
  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,
```

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```
legend_text = 8,
      dim_background_color = "white",
      spat_background_color = "white",
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      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
    dim_point_shape
                     dim reduction points with border or not (border or no_border)
   dim_point_size dim reduction plot: point size
    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_shape
                     spatial points with border or not (border or no_border)
    spat_point_size
```

spatial plot: point size

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```
spat_point_border_col
                  color of border around points
spat\_point\_border\_stroke
                  stroke size of border around points
                  size of point (cell)
midpoint
                  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_legend
                  show legend
legend_text
                  size of legend text
dim_background_color
                  color of plot background for dimension plot
spat_background_color
                  color of plot background for spatial plot
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plots
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

# See Also

spatDimGenePlot3D

# Examples

```
spatDimGenePlot2D(gobject)
```

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spatDimGenePlot3D

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,
  show_other_cells = T,
 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,
```

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```
z_ticks = NULL,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimGenePlot3D"
    )
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
    genes
                     genes to show
    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
    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
    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
                     size of point (cell)
    point_size
```

show\_legend

show legend

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#### **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_shape = c("border", "no_border"),
 dim_point_size = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border"),
  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,
```

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```
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,
  legend_symbol_size = 1,
  dim_background_color = "white",
  spat_background_color = "white",
 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
{\tt 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
                  = spatial dimension to use on x-axis
sdimx
sdimy
                  = spatial dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
```

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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 dim\_point\_shape point with border or not (border or no\_border) 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\_shape point with border or not (border or no\_border) 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  ${\tt dim\_center\_point\_size}$ 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  ${\tt spat\_center\_point\_size}$ 

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 legend\_symbol\_size size of legend symbols dim\_background\_color background color of points in dim. reduction space spat\_background\_color background color of spatial points size of axis text 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 242 spatDimPlot2D

#### **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_shape = c("border", "no_border"),
 dim_point_size = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border"),
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
```

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```
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,
  legend_symbol_size = 1,
 dim_background_color = "white",
  spat_background_color = "white",
 axis_text = 8,
 axis_title = 8,
  show_plot = NA,
 return_plot = NA,
  save_plot = NA,
 save_param = list(),
 default_save_name = "spatDimPlot2D"
)
```

### **Arguments**

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sdimx = spatial dimension to use on x-axis = spatial dimension to use on y-axis sdimy 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 subset of cells based on cell IDs select\_cells dim\_point\_shape point with border or not (border or no border) 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\_shape point with border or not (border or no\_border) 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 legend\_symbol\_size size of legend symbols dim\_background\_color background color of points in dim. reduction space spat\_background\_color background color of spatial points

size of axis text

axis\_text

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```
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,
```

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```
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
   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

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network\_name name of NN network to use, if show NN network = TRUE 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 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 dim\_point\_border\_col border color of points in dim. reduction space

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#### **Details**

Description of parameters.

#### Value

plotly

### **Examples**

```
spatDimPlot3D(gobject)
```

spatGenePlot

spatGenePlot

# **Description**

Visualize cells and gene expression according to spatial coordinates

```
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_shape = c("border", "no_border"),
 point_size = 1,
```

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```
point_border_col = "black",
      point_border_stroke = 0.1,
      show_legend = T,
      legend_text = 8,
      background_color = "white",
      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 = "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
    genes_low_color
                     color represents low gene expression
                     show underlying spatial network
    show_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
                     expression midpoint
   midpoint
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
   point_shape
                     point with border or not (border or no_border)
    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
```

spatGenePlot2D 251

```
legend_text
                  size of legend text
background_color
                  color of plot background
                  size of axis text
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
                  cowplot param: how to align
cow_align
show_plot
                  show plots
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
                  parameters for cowplot::save_plot()
```

#### **Details**

Description of parameters.

#### Value

ggplot

# See Also

spatGenePlot3D and spatGenePlot2D

# **Examples**

spatGenePlot(gobject)

| spatGenePlot2D | spatGenePlot2D |
|----------------|----------------|
|                |                |

# Description

Visualize cells and gene expression according to spatial coordinates

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#### Usage

network\_color

```
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_shape = c("border", "no_border"),
     point_size = 1,
     point_border_col = "black",
     point_border_stroke = 0.1,
     show_legend = T,
     legend_text = 8,
     background_color = "white",
     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 = "spatGenePlot2D"
   )
Arguments
   gobject
                    giotto object
   expression_values
                    gene expression values to use
                    genes to show
   genes
   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
```

color of spatial network

spatGenePlot2D 253

```
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 with border or not (border or no_border)
point_shape
                  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
legend_text
                  size of legend text
background_color
                  color of plot background
                  size of axis text
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 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

spatGenePlot3D

## **Examples**

spatGenePlot2D(gobject)

254 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
```

spatialAEH 255

```
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
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)

spatialAEH spatialAEH

## Description

Compute spatial variable genes with spatialDE method

256 spatialDE

#### Usage

```
spatialAEH(
  gobject = NULL,
  SpatialDE_results = NULL,
  name_pattern = "AEH_patterns",
  expression_values = c("raw", "normalized", "scaled", "custom"),
  pattern_num = 6,
  l = 1.05,
  python_path = NULL,
  return_gobject = TRUE
)
```

### **Arguments**

### **Details**

This function is a wrapper for the SpatialAEH method implemented in the ...

## Value

An updated giotto object

### **Examples**

```
spatialAEH(gobject)
```

spatialDE spatialDE

## Description

Compute spatial variable genes with spatialDE method

spatialDE 257

#### Usage

```
spatialDE(
  gobject = NULL,
  expression_values = c("raw", "normalized", "scaled", "custom"),
  size = c(4, 2, 1),
  color = c("blue", "green", "red"),
  sig_alpha = 0.5,
  unsig_alpha = 0.5,
  python_path = NULL,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "SpatialDE"
)
```

### **Arguments**

Giotto object gobject expression\_values gene expression values to use size size of plot color low/medium/high color scheme for plot sig\_alpha alpha value for significance unsig\_alpha alpha value for unsignificance specify specific path to python if required python\_path 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**

This function is a wrapper for the SpatialDE method implemented in the ...

#### Value

```
a list of data.frames with results and plot (optional)
```

```
spatialDE(gobject)
```

258 Spatial\_AEH

Spatial\_AEH

Spatial\_AEH

## Description

calculate automatic expression histology with spatialDE method

## 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

```
Spatial_AEH(gobject)
```

Spatial\_DE 259

Spatial\_DE

Spatial\_DE

# Description

calculate spatial varible genes with spatialDE method

## 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 specify specific path to python if required

# Details

Description.

### Value

a list or a dataframe of SVs

# **Examples**

```
Spatial_DE(gobject)
```

 ${\tt spatNetwDistributions} \ \textit{spatNetwDistributionsDistance}$ 

# Description

This function return histograms displaying the distance distribution for each spatial k-neighbor

260 spatNetwDistributions

#### Usage

```
spatNetwDistributions(
  gobject,
  spatial_network_name = "spatial_network",
  distribution = c("distance", "k_neighbors"),
  hist_bins = 30,
  test_distance_limit = NULL,
  ncol = 1,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "spatNetwDistributions"
)
```

#### **Arguments**

```
gobject
                  Giotto object
spatial_network_name
                  name of spatial network
distribution
                  show the distribution of cell-to-cell distance or number of k neighbors
hist_bins
                  number of binds to use for the histogram
test_distance_limit
                  effect of different distance threshold on k-neighbors
                  number of columns to visualize the histograms in
ncol
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, alternatively change save_name in save_param
```

#### **Details**

The **distance** option shows the spatial distance distribution for each nearest neighbor rank (1st, 2nd, 3th, ... neigbor). With this option the user can also test the effect of a distance limit on the spatial network. This distance limit can be used to remove neigbor cells that are considered to far away. The **k\_neighbors** option shows the number of k neighbors distribution over all cells.

### Value

ggplot plot

```
spatNetwDistributionsDistance(gobject)
```

```
spat {\tt NetwDistributionsDistance} \\ spat {\tt NetwDistributionsDistance}
```

## Description

This function return histograms displaying the distance distribution for each spatial k-neighbor

# Usage

```
spatNetwDistributionsDistance(
  gobject,
  spatial_network_name = "spatial_network",
  hist_bins = 30,
  test_distance_limit = NULL,
  ncol = 1,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "spatNetwDistributionsDistance")
```

### **Arguments**

```
gobject
                  Giotto object
spatial_network_name
                  name of spatial network
hist_bins
                  number of binds to use for the histogram
test_distance_limit
                  effect of different distance threshold on k-neighbors
                  number of columns to visualize the histograms in
ncol
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, alternatively change save_name in save_param
```

# Value

ggplot plot

```
spatNetwDistributionsDistance(gobject)
```

```
spat {\tt NetwDistributions} Kneighbors \\ spat {\tt NetwDistributions} Kneighbors
```

### **Description**

This function returns a histogram displaying the number of k-neighbors distribution for each cell

### Usage

```
spatNetwDistributionsKneighbors(
  gobject,
  spatial_network_name = "spatial_network",
  hist_bins = 30,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "spatNetwDistributionsKneighbors")
```

### **Arguments**

```
gobject
                  Giotto object
spatial_network_name
                  name of spatial network
hist_bins
                  number of binds to use for the histogram
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, alternatively change save_name in save_param
```

#### Value

ggplot plot

```
spatNetwDistributionsKneighbors(gobject)
```

spatPlot 263

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,
 select_cell_groups = NULL,
  select_cells = NULL,
 point_shape = c("border", "no_border"),
 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,
  legend_symbol_size = 1,
```

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```
background_color = "white",
      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
    gobject
                     giotto object
    group_by_subset
                     subset the group_by factor column
                     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_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_shape
                     point with border or not (border or no_border)
    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

spatPlot 265

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 legend\_symbol\_size size of legend symbols background\_color color of plot background 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 cowplot param: relative width cow\_rel\_w cow\_align cowplot param: how to 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.

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#### 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,
 cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
 gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
 point_shape = c("border", "no_border"),
 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,
```

spatPlot2D 267

```
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,
      legend_symbol_size = 1,
      background_color = "white",
      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
    gobject
                     giotto object
    group_by_subset
                     subset the group_by factor column
    sdimx
                     x-axis dimension name (default = 'sdimx')
                     y-axis dimension name (default = 'sdimy')
    sdimy
    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
```

point with border or not (border or no\_border)

size of point (cell)

point\_shape
point\_size

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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 underlying spatial network show\_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 legend\_symbol\_size size of legend symbols background\_color color of plot background axis\_text size of axis text axis\_title size of axis title cowplot param: how many columns cow\_n\_col cowplot param: relative height cow\_rel\_h

cowplot param: relative width

cowplot param: how to align

show plot

cow\_rel\_w
cow\_align

show\_plot

spatPlot2D\_single 269

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

```
spatPlot3D
```

#### **Examples**

```
spatPlot2D(gobject)
```

### **Description**

Visualize cells according to spatial coordinates

```
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_shape = c("border", "no_border"),
 point_size = 3,
 point_border_col = "black",
 point_border_stroke = 0.1,
 show_cluster_center = F,
  show_center_label = F,
```

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```
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,
  legend_symbol_size = 1,
 background_color = "white",
 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')
                  y-axis dimension name (default = 'sdimy')
sdimy
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 subset of cells based on cell IDs
select_cells
```

point\_shape point with border or not (border or no border) 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\_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 title of plot show legend show\_legend legend\_text size of legend text legend\_symbol\_size size of legend symbols background\_color color of plot background axis\_text size of axis text size of axis title 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 272 spatPlot3D

#### **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,
  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,
```

spatPlot3D 273

```
z_ticks = NULL,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spat3D"
    )
Arguments
    gobject
                      giotto object
    sdimx
                      x-axis dimension name (default = 'sdimx')
    sdimy
                      y-axis dimension name (default = 'sdimy')
                      z-axis dimension name (default = 'sdimy')
    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 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
                      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
```

title title of plot
show\_legend show legend
axis\_scale the way to scale the axis

custom\_ratio customize the scale of the plot
x\_ticks set the number of ticks on the x-axis
y\_ticks set the number of ticks on the y-axis
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]

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

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### **Details**

Description of parameters.

# Value

ggplot

# **Examples**

spatPlot3D(gobject)

spat\_fish\_func

spat\_fish\_func

# Description

performs fisher exact test

# Usage

```
spat_fish_func(gene, bin_matrix, spat_mat)
```

spat\_OR\_func

 $spat\_OR\_func$ 

# Description

calculate odds-ratio

```
spat_OR_func(gene, bin_matrix, spat_mat)
```

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

# Description

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

## Usage

```
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
                  second cell type
cell_type_2
                  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.

276 stitchFieldCoordinates

#### Value

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

### **Examples**

```
specificCellCellcommunicationScores(gobject)
```

#### **Description**

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

#### Usage

```
split_dendrogram_in_two(dend)
```

#### **Arguments**

dend

dendrogram object

#### Value

list of two dendrograms and height of node

#### **Examples**

```
{\tt split\_dendrogram\_in\_two(dend)}
```

```
stitchFieldCoordinates
```

stitchFieldCoordinates

## Description

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

```
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
)
```

subClusterCells 277

#### **Arguments**

```
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?
```

#### **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

278 subClusterCells

```
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
    )
Arguments
    gobject
                     giotto object
    name
                     name for new clustering result
    cluster_method clustering method to use
    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
                     parameters for parameters for createNearestNetwork
    nn_param
                     number of k for createNearestNetwork
    k_neighbors
                     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)
    network_name
                     name of NN network to use
    return_gobject boolean: return giotto object (default = TRUE)
```

verbose

verbose

subsetGiotto 279

#### **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

subsetGiotto

## 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

```
subsetGiotto(gobject)
```

280 subsetGiottoLocs

subsetGiottoLocs

subsetGiottoLocs

## Description

subsets Giotto object based on spatial locations

## Usage

```
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
)
```

## **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
```

```
subsetGiottoLocs(gobject)
```

trendSceek 281

|--|

### **Description**

Compute spatial variable genes with trendsceek method

### Usage

```
trendSceek(
  gobject,
  expression_values = c("normalized", "raw"),
  subset_genes = NULL,
  nrand = 100,
  ncores = 8,
  ...
)
```

## **Arguments**

### **Details**

This function is a wrapper for the trendsceek\_test method implemented in the trendsceek package

### Value

data.frame with trendsceek spatial genes results

```
trendSceek(gobject)
```

282 viewHMRFresults

viewHMRFresults

viewHMRFresults

# Description

View results from doHMRF.

## Usage

```
viewHMRFresults(
  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

```
visPlot
```

```
viewHMRFresults(gobject)
```

viewHMRFresults2D 283

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

number of fivire 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

```
spatPlot2D
```

```
viewHMRFresults2D(gobject)
```

284 viewHMRFresults3D

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
```

```
viewHMRFresults3D(gobject)
```

violinPlot 285

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"
)
```

# Arguments

```
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
```

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#### 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
)
```

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#### **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)

```
vis {\tt DimGenePlot\_2D\_ggplot} \\ vis {\tt DimGenePlot\_2D\_ggplot}
```

### **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
```

### **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
                 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.

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#### 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,
```

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```
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

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#### **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
                     color for cells (see details)
    cell_color
    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 295

#### **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**

```
\verb|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
```

# **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
                 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)
```

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```
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,
```

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```
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**

```
giotto object
gobject
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
```

300 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
```

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#### **Arguments**

gobject giotto object expression\_values gene expression values to use genes to show genes 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 underlying spatial network show\_network network\_color color of spatial network spatial\_network\_name name of spatial network to use show spatial grid show\_grid color of spatial grid grid\_color 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 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

two methods of plot

show plots

Details

z\_ticks
plot\_method

show\_plots

Description of parameters.

#### Value

ggplot or plotly

## **Examples**

```
visGenePlot(gobject)
```

```
{\tt visGenePlot\_2D\_ggplot} \quad {\it 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
)
```

# 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
```

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

genes\_high\_color

genes\_mid\_color

genes\_low\_color

spatial\_grid\_name

point\_size show\_legend

axis\_scale

x\_ticks

y\_ticks

```
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
Arguments
   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
```

color represents high gene expression

color represents middle gene expression

color represents low gene expression

name of spatial grid to use

three mode to adjust axis scale

number of ticks on x axis number of ticks on y axis

size of point (cell)

show legend

visPlot 305

#### **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,
```

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```
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,
)
```

## **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)
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 307

```
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,
```

308 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,
)
```

## **Arguments**

```
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 309

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**

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

310 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
```

## **Arguments**

```
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 311

```
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,
```

312 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,
sticks = 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 313

#### **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,
```

314 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 to show
   genes
   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
```

visSpatDimGenePlot 315

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)

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

 ${\tt edge\_alpha\_dim} \ \ {\tt dim} \ {\tt reduction} \ {\tt plot}; \ {\tt column} \ {\tt to} \ {\tt use} \ {\tt for} \ {\tt alpha} \ {\tt of} \ {\tt the} \ {\tt edges}$ 

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
```

show\_spatial\_grid

spatial\_network\_alpha

id

alpha of spatial network

sh

show spatial grid

network\_color color of spatial/nn network

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

## **Details**

Description of parameters.

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#### 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,
```

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```
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
```

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```
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

 $show_NN_network = F,$ 

network\_name
cell\_color

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```
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
   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

color for cells (see details)

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```
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 325

visSpatDimPlot\_3D
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

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```
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 327

## **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 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

```
write\_giotto\_viewer\_numeric\_annotation \\ write\_giotto\_viewer\_numeric\_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

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