# Package 'Giotto'

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```
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      matrixStats (>= 0.55.0),
      IRanges,
      devtools,
      reshape2,
      ggraph,
```

2 R topics documented:

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Rcpp,
     rlang (>= 0.4.3),
     fit distrplus\\
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
```

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addCellIntMetadata addC

addCellIntMetadata

#### **Description**

Creates an additional metadata column with information about interacting and non-interacting cell types of the selected cell-cell interaction.

## Usage

```
addCellIntMetadata(
  gobject,
  spatial_network = "spatial_network",
  cluster_column,
  cell_interaction,
  name = "select_int",
  return_gobject = TRUE
)
```

#### **Arguments**

```
gobject giotto object

spatial_network

name of spatial network to use

cluster_column column of cell types

cell_interaction

cell-cell interaction to use

name name for the new metadata column

return_gobject return an updated giotto object
```

#### **Details**

This function will create an additional metadata column which selects interacting cell types for a specific cell-cell interaction. For example, if you want to color interacting astrocytes and oligodendrocytes it will create a new metadata column with the values "select\_astrocytes", "select\_oligodendrocytes", "other\_astrocytes", "other\_oligodendrocytes" and "other". Where "other" is all other cell types found within the selected cell type column.

# Value

Giotto object

## **Examples**

```
addCellIntMetadata(gobject)
```

addCellMetadata 9

addCellMetadata addCellMetadata

## **Description**

adds cell metadata to the giotto object

#### Usage

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

# Arguments

gobject giotto object

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

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

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

## **Details**

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

#### Value

giotto object

# **Examples**

addCellMetadata(gobject)

addCellStatistics addCellStatistics

# Description

adds cells statistics to the giotto object

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#### Usage

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

# Arguments

## **Details**

This function will add the following statistics to cell metadata:

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

#### Value

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

## **Examples**

```
addCellStatistics(gobject)
```

addGeneMetadata

addGeneMetadata

#### **Description**

adds gene metadata to the giotto object

## Usage

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

# 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
```

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#### **Details**

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

#### Value

giotto object

# **Examples**

addGeneMetadata(gobject)

addGeneStatistics

addGeneStatistics

#### **Description**

adds gene statistics to the giotto object

#### Usage

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

## **Arguments**

#### **Details**

This function will add the following statistics to gene metadata:

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

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#### Value

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

# **Examples**

addGeneStatistics(gobject)

addHMRF

addHMRF

# Description

Add selected results from doHMRF to the giotto object

# Usage

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

# Arguments

gobject giotto object

HMRF output from doHMRF()

k number of domains

name specify a custom name

# **Details**

Description ...

#### Value

giotto object

# **Examples**

addHMRF(gobject)

addNetworkLayout 13

addNetworkLayout

addNetworkLayout

# Description

Add a network layout for a selected nearest neighbor network

## Usage

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

## Arguments

## **Details**

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

## Value

giotto object with updated layout for selected NN network

# Examples

```
addNetworkLayout(gobject)
```

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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)
```

# Description

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

#### Usage

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

```
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)
```

# 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

## Usage

```
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

# Usage

```
binGetSpatialGenes(
  gobject,
  bin_method = c("kmeans", "rank"),
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_network_name = "spatial_network",
  nstart = 3,
  iter_max = 10,
  percentage_rank = 10,
  do_fisher_test = TRUE,
  calc_hub = FALSE,
  hub_min_int = 3,
  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

calc\_hub calculate the number of hub cells

hub\_min\_int minimum number of cell-cell interactions for a hub cell

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

## Usage

```
calculateHVG(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  method = c("cov_groups", "cov_loess"),
  reverse_log_scale = FALSE,
  logbase = 2,
  expression_threshold = 0,
  nr_expression_groups = 20,
  zscore_threshold = 1.5,
  HVGname = "hvg",
  difference_in_cov = 0.1,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "HVGplot",
  return_gobject = TRUE
)
```

24 calculateHVG

#### **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)

calculateMetaTable 25

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
```

calculate Meta Table Cells

# Description

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

## Usage

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

#### **Arguments**

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

#### Value

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

## **Examples**

```
calculateMetaTableCells(gobject)
```

#### **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.

cellProximityBarplot 27

#### 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 29

```
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
```

#### **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**

```
giotto object
gobject
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
```

cellProximitySpatPlot 31

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

## **Details**

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

#### Value

igraph plot

# **Examples**

```
cellProximityNetwork(CPscore)
```

```
cellProximitySpatPlot cellProximitySpatPlot
```

## **Description**

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

# Usage

```
{\tt cellProximitySpatPlot(gobject,\ \ldots)}
```

# 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_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

 $cell Proximity SpatPlot 2D \ and \ cell Proximity SpatPlot 3D \ for \ 3D$ 

#### **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
\verb"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

## Usage

```
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)

cellProximityVisPlot 37

```
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"),
)
```

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)

#### **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,
)
```

```
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,
)
```

```
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,
```

```
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)
```

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()

clusterCells 45

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,
```

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```
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

 $\verb"pyth_leid_iterations"$ 

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

clusterSpatialCorGenes

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```
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
)
```

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#### **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)
```

combCCcom

#### **Description**

Combine spatial and expression based cell-cell communication data.tables

combCCcom

# Usage

```
combCCcom(
  spatialCC,
  exprCC,
  min_lig_nr = 3,
  min_rec_nr = 3,
  min_padj_value = 1,
  min_log2fc = 0,
  min_av_diff = 0
)
```

## **Arguments**

```
spatialCC spatial cell-cell communication scores
exprCC expression cell-cell communication scores
min_lig_nr minimum number of ligand cells
min_rec_nr minimum number of receptor cells
min_padj_value minimum adjusted p-value
min_log2fc minimum log2 fold-change
min_av_diff minimum average expression difference
```

## Value

combined data.table with spatial and expression communication data

```
combCCcom(gobject)
```

```
combineCellProximityGenes
```

combineCellProximityGenes

# Description

Combine CPG scores in a pairwise manner.

# Usage

```
combineCellProximityGenes(
  cpgObject,
  selected_ints = NULL,
  selected_genes = NULL,
  specific_genes_1 = NULL,
  specific_genes_2 = NULL,
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0,
  min_log2_fc = 0.5,
  do_parallel = TRUE,
  cores = NA,
  verbose = T
)
```

# **Arguments**

```
cpgObject
                  cell proximity gene score object
selected_ints
                  subset of selected cell-cell interactions (optional)
selected_genes subset of selected genes (optional)
specific_genes_1
                  specific geneset combo (need to position match specific_genes_2)
specific_genes_2
                  specific geneset combo (need to position match specific_genes_1)
min_cells
                  minimum number of target cell type
min_int_cells
                  minimum number of interacting cell type
min_fdr
                  minimum adjusted p-value
                  minimum absolute spatial expression difference
min_spat_diff
min_log2_fc
                  minimum absolute log2 fold-change
do_parallel
                  run calculations in parallel with mclapply
                  number of cores to use if do_parallel = TRUE
cores
                  verbose
verbose
```

#### Value

cpgObject that contains the filtered differential gene scores

50 combineCPG

## **Examples**

```
combineCellProximityGenes(gobject)
```

```
combine \verb|CellProximityGenes_per_interaction| \\ combine CellProximity Genes\_per\_interaction|
```

## **Description**

Combine CPG scores per interaction

## Usage

```
combineCellProximityGenes_per_interaction(
  cpgObject,
  sel_int,
  selected_genes = NULL,
  specific_genes_1 = NULL,
  specific_genes_2 = NULL,
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0,
  min_log2_fc = 0.5
)
```

## **Examples**

combineCellProximityGenes\_per\_interaction()

combineCPG

combineCPG

# Description

Combine CPG scores in a pairwise manner.

## Usage

```
combineCPG(
  cpgObject,
  selected_ints = NULL,
  selected_genes = NULL,
  specific_genes_1 = NULL,
  specific_genes_2 = NULL,
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0,
```

combineMetadata 51

```
min_log2_fc = 0.5,
  do_parallel = TRUE,
  cores = NA,
  verbose = T
)
```

## **Arguments**

```
cpg0bject
                  cell proximity gene score object
                  subset of selected cell-cell interactions (optional)
selected_ints
selected_genes subset of selected genes (optional)
specific_genes_1
                  specific geneset combo (need to position match specific_genes_2)
specific_genes_2
                  specific geneset combo (need to position match specific_genes_1)
min_cells
                  minimum number of target cell type
min_int_cells minimum number of interacting cell type
min_fdr
                  minimum adjusted p-value
min_spat_diff
                  minimum absolute spatial expression difference
                  minimum absolute log2 fold-change
min_log2_fc
                  run calculations in parallel with mclapply
do_parallel
                  number of cores to use if do_parallel = TRUE
cores
verbose
                  verbose
```

### Value

cpgObject that contains the filtered differential gene scores

#### **Examples**

```
combineCPG(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)
```

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

52 combine\_ints\_f

#### Value

Extended cell metadata in data.table format.

## **Examples**

```
combineMetadata(gobject)
```

```
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
                  log2 fold-change threshold
min_log2_fc
\min_{pval}
                  p-value threshold
```

#### Value

Gene to gene scores in data.table format

convertEnsemblToGeneSymbol

convert Ensembl To Gene Symbol

# Description

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

## Usage

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

## **Arguments**

matrix an expression matrix with ensembl gene IDs as rownames

species species to use for gene symbol conversion

## **Details**

This function requires that the biomaRt library is installed

## Value

expression matrix with gene symbols as rownames

# **Examples**

convertEnsemblToGeneSymbol(matrix)

# Description

convert to a full spatial network

# Usage

```
convert_to_full_spatial_network(reduced_spatial_network_DT)
```

```
convert\_to\_reduced\_spatial\_network\\ convert\_to\_reduced\_spatial\_network
```

## **Description**

convert to a reduced spatial network

#### Usage

```
convert_to_reduced_spatial_network(full_spatial_network_DT)
```

createDelaunayNetwork

## **Description**

Create a spatial Delaunay network based on cell centroid physical distances.

## Usage

```
createDelaunayNetwork(
  gobject,
  dimensions = c("sdimx", "sdimy"),
  name = "delaunay_network",
  maximum_distance = "auto",
  minimum_k = 0,
  Y = TRUE,
  j = TRUE,
  S = 0,
  verbose = T,
  return_gobject = TRUE,
  ...
)
```

## **Arguments**

gobject giotto object

dimensions which spatial dimensions to use (maximum 2 dimensions)

name name for spatial network (default = 'delaunay\_network')

maximum\_distance

distance cuttof for Delaunay neighbors to consider

minimum\_k minimum neighbours if maximum\_distance != NULL

Y (RTriangle) If TRUE prohibits the insertion of Steiner points on the mesh bound-

ary.

j (RTriangle) If TRUE jettisons vertices that are not part of the final triangulation

from the output.

createGiottoInstructions 55

```
S (RTriangle) Specifies the maximum number of added Steiner points.

verbose verbose

return_gobject boolean: return giotto object (default = TRUE)

... Other parameters of the triangulate function
```

#### **Details**

Creates a spatial Delaunay network as explained in triangulate.

#### Value

giotto object with updated spatial network slot

# **Examples**

```
createDelaunayNetwork(gobject)
```

```
createGiottoInstructions
```

*createGiottoInstructions* 

#### **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
```

```
path to python binary to use
python_path
show_plot
                  print plot to console, default = TRUE
return_plot
                  return plot as object, default = TRUE
                  automatically save plot, dafault = FALSE
save_plot
save_dir
                  path to directory where to save plots
                  resolution for raster images
dpi
height
                  height of plots
width
                  width of plots
```

56 createGiottoObject

#### 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
```

```
matrix with raw expression counts [required]
raw_exprs
                  data.table or data.frame with coordinates for cell centroids
spatial_locs
norm_expr
                  normalized expression values
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)
```

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```
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
offset_file
                  file used to stitch fields together (optional)
instructions
                  list of instructions or output result from createGiottoInstructions
```

#### Details

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

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

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

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

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

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

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

#### Value

giotto object

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

58 createHeatmap\_DT

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
cluster_order
                 method to determine 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
```

## **Details**

Creates input data.tables for plotHeatmap function.

createMetagenes 59

#### 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)
```

60 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,
  ...
)
```

```
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 61

#### **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.

62 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 63

### **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 65

```
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)
```

66 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 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")
)
```

```
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 69

```
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

70 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)
```

detectSpatialCorGenes 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
)
```

```
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 73

#### **Details**

Steps to identify spatial patterns:

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

### Value

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

## **Examples**

```
detectSpatialPatterns(gobject)
```

dimCellPlot

dimCellPlot

### **Description**

Visualize cells according to dimension reduction coordinates

```
dimCellPlot(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  cell_annotation_values = 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

dimCellPlot2D

### **Description**

Visualize cells according to dimension reduction coordinates

```
dimCellPlot2D(
  gobject,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  spat_enr_names = NULL,
  cell_annotation_values = 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,
```

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```
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

## **Examples**

```
dimCellPlot2D(gobject)
```

dimGenePlot 79

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 81

#### **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() . . .

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### **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)
                 name of NN network to use, if show_NN_network = TRUE
network_name
                 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 85

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,
```

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```
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 87

```
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)
```

88 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 89

```
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

90 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 91

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,
```

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```
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 93

```
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
```

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

### **Details**

Description of parameters. For 3D plots see dimPlot3D

## Value

ggplot

## **Examples**

dimPlot2D\_single(gobject)

dimPlot3D dimPlot3D

## **Description**

Visualize cells according to dimension reduction coordinates

94 dimPlot3D

### 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 95

```
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)

96 doHclust

## **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 97

```
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",
```

98 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
                 overwrite output folder
```

#### **Details**

Description of HMRF parameters ...

## Value

Creates a directory with results that can be viewed with viewHMRFresults

# **Examples**

```
doHMRF(gobject)
```

doKmeans 99

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
```

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### **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 101

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

doLeidenSubCluster

## **Description**

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

102 doLeidenSubCluster

#### **Usage**

```
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
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
```

name of NN network to use

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

verbose

network\_name

verbose

doLouvainCluster 103

### **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 107

 ${\tt doLouvainSubCluster} \qquad {\tt doLouvainSubCluster}$ 

### **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_community} 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
```

112 doRandomWalkCluster

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

doSNNCluster 113

#### **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_cell_proximity_test
```

 $do\_cell\_proximity\_test$ 

## Description

Performs a selected differential test on subsets of a matrix

do\_limmatest 115

### **Examples**

```
do_cell_proximity_test()
```

do\_limmatest

do\_limmatest

## Description

Performs limma t.test on subsets of a matrix

## Usage

```
do_limmatest(expr_values, select_ind, other_ind)
```

## **Examples**

```
do_limmatest()
```

```
do_multi_permuttest_random
```

do\_multi\_permuttest\_random

## Description

calculate multiple random values

## Usage

```
do_multi_permuttest_random(
  expr_values,
  select_ind,
  other_ind,
  n = 100,
  cores = 2
)
```

```
{\tt do\_multi\_permuttest\_random()}
```

```
\begin{tabular}{ll} $do\_permuttest\_original \\ \hline & do\_permuttest\_original \\ \end{tabular}
```

## Description

calculate original values

## Usage

```
do_permuttest_original(expr_values, select_ind, other_ind, name = "orig")
```

## **Examples**

```
do_permuttest_original()
```

```
do_permuttest_random do_permuttest_random
```

# Description

calculate random values

Performs permutation test on subsets of a matrix

## Usage

```
do_permuttest_random(expr_values, select_ind, other_ind, name = "perm_1")
do_permuttest(
   expr_values,
   select_ind,
   other_ind,
   n_perm = 100,
   adjust_method = "fdr",
   cores = 2
)
```

```
do_permuttest_random()
do_permuttest_random()
```

### **Description**

smooth gene expression over a defined spatial grid

#### Usage

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

## **Arguments**

#### Value

matrix with smoothened gene expression values based on spatial grid

### **Examples**

```
{\tt 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

118 do\_ttest

#### Usage

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

#### **Arguments**

```
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)
```

#### **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)
```

 $do\_ttest$ 

do\_ttest

#### **Description**

Performs t.test on subsets of a matrix

Performs wilcoxon on subsets of a matrix

```
do_ttest(expr_values, select_ind, other_ind, adjust_method)
do_wilctest(expr_values, select_ind, other_ind, adjust_method)
```

DT\_removeNA 119

## **Examples**

```
do_ttest()
do_ttest()
```

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

exportGiot to Viewer

## **Description**

compute highly variable genes

120 exportGiottoViewer

#### 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
    )
Arguments
                     giotto object
    gobject
    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
```

#### **Details**

overwrite\_dir

verbose

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.

numerical indicating how to round the expression data overwrite files in the directory if it already existed

#### Value

writes the necessary output to use in Giotto Viewer

be verbose

exprCellCellcom 121

#### **Examples**

```
exportGiottoViewer(gobject)
```

exprCellCellcom

exprCellCellcom

### **Description**

Cell-Cell communication scores based on expression only

### Usage

### **Arguments**

```
gobject
                  giotto object to use
cluster_column cluster column with cell type information
random_iter
                  number of iterations
                  first specific gene set from gene pairs
gene_set_1
gene_set_2
                  second specific gene set from gene pairs
log2FC\_addendum
                  addendum to add when calculating log2FC
                  which method to adjust p-values
adjust_method
adjust_target
                  adjust multiple hypotheses at the cell or gene level
verbose
                  verbose
```

#### Details

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

#### Value

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

```
exprCellCellcom(gobject)
```

122 extractNearestNetwork

```
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

```
extractNearestNetwork(gobject)
```

fDataDT 123

fDataDT

fDataDT

# Description

show gene metadata

## Usage

```
fDataDT(gobject)
```

## Arguments

gobject

giotto object

#### Value

data.table with gene metadata

## **Examples**

```
pDataDT(gobject)
```

```
filter {\tt CellProximityGenes}
```

filter Cell Proximity Genes

## Description

Filter cell proximity gene scores.

```
filterCellProximityGenes(
  cpgObject,
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0.2,
  min_log2_fc = 0.5,
  direction = c("both", "up", "down")
```

124 filterCombinations

#### **Arguments**

```
cpgObject cell proximity gene score object
min_cells minimum number of target cell type
min_int_cells minimum number of interacting cell type
min_fdr minimum adjusted p-value
min_spat_diff minimum absolute spatial expression difference
min_log2_fc minimum absolute log2 fold-change
direction differential expression directions to keep
```

#### Value

cpgObject that contains the filtered differential gene scores

#### **Examples**

```
filterCellProximityGenes(gobject)
```

filterCombinations filterCombinations

### **Description**

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

### Usage

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

## Arguments

```
gobject giotto object

expression_values

expression values to use

expression_thresholds

all thresholds to consider a gene expressed

gene_det_in_min_cells

minimum number of cells that should express a gene to consider that gene further
```

filterCPG 125

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

filterCPG filterCPG

#### **Description**

Filter cell proximity gene scores.

## Usage

```
filterCPG(
  cpgObject,
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0.2,
  min_log2_fc = 0.5,
  direction = c("both", "up", "down")
)
```

## Arguments

```
cpgObject cell proximity gene score object
min_cells minimum number of target cell type
min_int_cells minimum number of interacting cell type
min_fdr minimum adjusted p-value
min_spat_diff minimum absolute spatial expression difference
min_log2_fc minimum absolute log2 fold-change
direction differential expression directions to keep
```

126 filterCPGscores

#### Value

cpgObject that contains the filtered differential gene scores

#### **Examples**

```
filterCPG(gobject)
```

filterCPGscores

filterCPGscores

#### **Description**

visualize Cell Proximity Gene enrichment scores

### Usage

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

#### **Arguments**

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

### **Details**

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

#### Value

Gene to gene scores in data.table format

```
filterCPGscores(CPGscore)
```

filterDistributions 127

```
filterDistributions filterDistributions
```

#### **Description**

show gene or cell distribution after filtering on expression threshold

## Usage

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

## Arguments

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

### Value

ggplot object

```
filterDistributions(gobject)
```

128 filterGiotto

filterGiotto

filterGiotto

## Description

filter Giotto object based on expression threshold

#### Usage

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

#### **Arguments**

```
gobject giotto object

expression_values

expression values to use

expression_threshold

threshold to consider a gene expressed

gene_det_in_min_cells

minimum # of cells that need to express a gene

min_det_genes_per_cell

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

verbose

verbose
```

#### **Details**

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

#### Value

giotto object

```
filterGiotto(gobject)
```

```
findCellProximityGenes
```

findCellProximityGenes

## Description

Identifies genes that are differentially expressed due to proximity to other cell types.

#### Usage

### **Arguments**

```
gobject
                 giotto object
expression_values
                 expression values to use
cluster_column name of column to use for cell types
spatial_network_name
                 name of spatial network to use
minimum_unique_cells
                 minimum number of target cells required
minimum_unique_int_cells
                 minimum number of interacting cells required
                 which differential expression test
diff_test
adjust_method which method to adjust p-values
nr_permutations
                 number of permutations if diff_test = permutation
exclude_selected_cells_from_test
                 exclude interacting cells other cells
                 run calculations in parallel with mclapply
do_parallel
                 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 (approximated by physical proximity) with other cell types. The results data.table in the cpgObject contains - at least - the following columns:

- genes: All or selected list of tested genes
- sel: average gene expression in the interacting cells from the target cell type
- other: average gene expression in the NOT-interacting cells from the target cell type
- log2fc: log2 fold-change between sel and other
- diff: spatial expression difference between sel and other
- p.value: associated p-value
- p.adj: adjusted p-value
- cell\_type: target cell type
- int\_cell\_type: interacting cell type
- nr\_select: number of cells for selected target cell type
- int\_nr\_select: number of cells for interacting cell type
- nr\_other: number of other cells of selected target cell type
- int\_nr\_other: number of other cells for interacting cell type
- unif int: cell-cell interaction

#### Value

cpgObject that contains the differential gene scores

#### **Examples**

```
findCellProximityGenes(gobject)
```

```
find {\tt CellProximityGenes\_per\_interaction} \\ find {\tt CellProximityGenes\_per\_interaction}
```

### **Description**

Identifies genes that are differentially expressed due to proximity to other cell types.

```
findCellProximityGenes_per_interaction(
   expr_values,
   cell_metadata,
   annot_spatnetwork,
   sel_int,
   minimum_unique_cells = 1,
   minimum_unique_int_cells = 1,
   exclude_selected_cells_from_test = T,
   diff_test = c("permutation", "limma", "t.test", "wilcox"),
   adjust_method = "bonferroni",
   nr_permutations = 100,
   cores = 1
)
```

findCPG 131

#### **Examples**

findCellProximityGenes\_per\_interaction()

findCPG

findCPG

#### **Description**

Identifies genes that are differentially expressed due to proximity to other cell types.

## Usage

## **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
cluster_column name of column to use for cell types
spatial_network_name
                  name of spatial network to use
minimum_unique_cells
                  minimum number of target cells required
minimum_unique_int_cells
                  minimum number of interacting cells required
diff_test
                  which differential expression test
                  which method to adjust p-values
adjust_method
nr_permutations
                  number of permutations if diff_test = permutation
{\tt exclude\_selected\_cells\_from\_test}
                  exclude interacting cells other cells
do_parallel
                  run calculations in parallel with mclapply
                  number of cores to use if do_parallel = TRUE
cores
```

132 findGiniMarkers

#### **Details**

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

- genes: All or selected list of tested genes
- sel: average gene expression in the interacting cells from the target cell type
- other: average gene expression in the NOT-interacting cells from the target cell type
- log2fc: log2 fold-change between sel and other
- diff: spatial expression difference between sel and other
- p.value: associated p-value
- p.adj: adjusted p-value
- cell\_type: target cell type
- int\_cell\_type: interacting cell type
- nr\_select: number of cells for selected target cell type
- int\_nr\_select: number of cells for interacting cell type
- nr\_other: number of other cells of selected target cell type
- int\_nr\_other: number of other cells for interacting cell type
- unif\_int: cell-cell interaction

#### Value

cpgObject that contains the differential gene scores

### **Examples**

```
findCPG(gobject)
```

findGiniMarkers

findGiniMarkers

### **Description**

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

```
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
)
```

findGiniMarkers 133

#### **Arguments**

```
giotto object
gobject
expression_values
                  gene expression values to use
cluster_column clusters to use
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
                  filter on minimum gini coefficient for expression
min_det_gini_score
                  filter on minimum gini coefficient for detection
detection_threshold
                  detection threshold for gene expression
                  rank scores for both detection and expression to include
rank_score
                  minimum number of top genes to return
min_genes
```

#### **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
                  minimum number of top genes to return
min_genes
verbose
                  be verbose
```

#### Value

data.table with marker genes

#### See Also

findGiniMarkers

findMarkers 135

#### **Examples**

```
findGiniMarkers_one_vs_all(gobject)
```

findMarkers

findMarkers

## Description

Identify marker genes for selected clusters.

### Usage

```
findMarkers(
 gobject,
  expression_values = c("normalized", "scaled", "custom"),
 cluster_column,
 method = c("scran", "gini", "mast"),
 subset_clusters = NULL,
 group_1 = NULL,
 group_2 = NULL,
 min_expr_gini_score = 0.5,
 min_det_gini_score = 0.5,
 detection_threshold = 0,
 rank_score = 1,
 min\_genes = 4,
 group_1_name = NULL,
 group_2_name = NULL,
 adjust_columns = NULL,
```

### **Arguments**

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
method
                  method to use to detect differentially expressed genes
subset_clusters
                  selection of clusters to compare
group_1
                  group 1 cluster IDs from cluster_column for pairwise comparison
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
min_expr_gini_score
                  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
min_genes minimum number of top genes to return (for gini)
group_1_name mast: custom name for group_1 clusters
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 137

#### **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)
```

138 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)
```

#### **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)
```

140 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

## **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)
```

```
\label{lem:cone_vs_all} find Scran Markers\_one\_vs\_all \\ find Scran Markers\_one\_vs\_all
```

## Description

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

#### Usage

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

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

### Value

data.table with marker genes

## See Also

findScranMarkers

```
findScranMarkers_one_vs_all(gobject)
```

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)
```

144 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 145

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)
```

146 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 147

get10Xmatrix

get10Xmatrix

## **Description**

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

# Usage

```
get10Xmatrix(path_to_data, gene_column_index = 1)
```

### **Arguments**

```
path_to_data path to the 10X folder gene_column_index which column from the features or genes .tsv file to use for row ids
```

### **Details**

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

- barcodes.tsv(.gz)
- features.tsv(.gz) or genes.tsv(.gz)
- matrix.mtx(.gz)

By default the first column of the features or genes .tsv file will be used, however if multiple annotations are provided (e.g. ensembl gene ids and gene symbols) the user can select another column.

#### Value

expression matrix from 10X

# **Examples**

```
get10Xmatrix(10Xmatrix)
```

```
{\it get Cell Proximity Gene Scores} \\ {\it get Cell Proximity Gene Scores}
```

### **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)

150 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 151

### 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**

```
gobject giotto object
expression_values
expression values to use

cluster_column name of column to use for clusters

cor correlation score to calculate distance

distance distance method to use for hierarchical clustering

h height of horizontal lines to plot

h_color color of horizontal lines

show_dend show dendrogram
```

#### **Details**

verbose

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)
```

be verbose

```
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 155

```
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
```

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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.

#### 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 159

### **Examples**

```
hyperGeometricEnrich(gobject)
```

kmeans\_binarize

kmeans binarize

### **Description**

create binarized scores using kmeans

#### Usage

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

loadHMRF

loadHMRF

# **Description**

load previous HMRF

### Usage

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

# Arguments

```
name of HMRF that was run
name_used
output_folder_used
                 output folder that was used
                 number of HMRF domains that was tested
k_used
```

betas\_used betas that were tested python\_path\_used

python path that was used

### **Details**

Description of HMRF parameters ...

### Value

reloads a previous ran HMRF from doHRMF

```
loadHMRF(gobject)
```

160 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.

```
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,
```

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```
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 163

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)
```

164 normalizeGiotto

normalizeGiotto

normalize Giotto

#### **Description**

fast 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,
  log_offset = 1,
  logbase = 2,
  scale_genes = T,
  scale_cells = T,
  scale_order = c("first_genes", "first_cells"),
  verbose = F
)
```

### **Arguments**

```
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
                  transform values to log-scale
log_norm
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
```

### **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.

normalizeGiottoOld 165

- B. The normalization method as provided by the osmFISH paper is also implemented:
  - 1. First normalize genes, for each gene divide the counts by the total gene count and multiply by the total number of genes.
  - 2. Next normalize cells, for each cell divide the normalized gene counts by the total counts per cell and multiply by the total number of cells.

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

#### Value

giotto object

# **Examples**

```
normalizeGiotto(gobject)
```

normalizeGiottoOld

normalizeGiotto

#### **Description**

normalize and/or scale expresion values of Giotto object

#### Usage

```
normalizeGiottoOld(
  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
)
```

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

OR\_function2

#### **Details**

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

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

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

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

#### Value

giotto object

### **Examples**

normalizeGiotto(gobject)

OR\_function2

OR\_function2

# Description

calculate odds-ratio

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

PAGEEnrich 167

**PAGEEnrich** 

**PAGEEnrich** 

#### **Description**

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

#### Usage

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

## **Arguments**

```
gobject Giotto object

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

expression values to use

reverse_log_scale

reverse expression values from log scale

logbase log base to use if reverse_log_scale = TRUE

output_enrichment

how to return enrichment output
```

### **Details**

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

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

#### Value

data.table with enrichment results

### See Also

makeSignMatrixPAGE

```
PAGEEnrich(gobject)
```

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

```
pDataDT(gobject)
```

plotCCcomDotplot 169

 $plot {\tt CCcomDotplot} \\ plot {\tt CCcomDotplot}$ 

#### **Description**

Plots dotplot for ligand-receptor communication scores in cell-cell interactions

#### Usage

```
plotCCcomDotplot(
  gobject,
  comScores,
  selected_LR = NULL,
  selected_cell_LR = NULL,
  show_LR_names = TRUE,
  show_cell_LR_names = TRUE,
  cluster_on = c("PI", "LR_expr", "log2fc"),
  cor_method = c("pearson", "kendall", "spearman"),
  aggl_method = c("ward.D", "ward.D2", "single", "complete", "average", "mcquitty",
    "median", "centroid"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCCcomDotplot"
)
```

```
giotto object
gobject
comScores
                  communinication scores from exprCellCellcom or spatCellCellcom
                  selected ligand-receptor combinations
selected_LR
selected_cell_LR
                  selected cell-cell combinations for ligand-receptor combinations
show_LR_names
                  show ligand-receptor names
show_cell_LR_names
                  show cell-cell names
                  values to use for clustering of cell-cell and ligand-receptor pairs
cluster_on
                  correlation method used for clustering
cor\_method
aggl_method
                  agglomeration method used by hclust
                  show plots
show_plot
return_plot
                  return plotting 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
                  values to show on heatmap
show
```

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#### Value

ggplot

#### **Examples**

```
plotCCcomDotplot(CPGscores)
```

plotCCcomHeatmap

plotCCcomHeatmap

## **Description**

Plots heatmap for ligand-receptor communication scores in cell-cell interactions

### Usage

```
plotCCcomHeatmap(
  gobject,
  comScores,
  selected_LR = NULL,
  selected_cell_LR = NULL,
  show_LR_names = TRUE,
  show_cell_LR_names = TRUE,
  show = c("PI", "LR_expr", "log2fc"),
 cor_method = c("pearson", "kendall", "spearman"),
aggl_method = c("ward.D", "ward.D2", "single", "complete", "average", "mcquitty",
     "median", "centroid"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCCcomHeatmap"
)
```

```
gobject
                 giotto object
comScores
                 communinication scores from exprCellCellcom or spatCellCellcom
selected_LR
                 selected ligand-receptor combinations
selected_cell_LR
                 selected cell-cell combinations for ligand-receptor combinations
                 show ligand-receptor names
show_LR_names
show_cell_LR_names
                 show cell-cell names
show
                 values to show on heatmap
cor_method
                 correlation method used for clustering
                 agglomeration method used by hclust
aggl_method
show_plot
                 show plots
```

#### Value

ggplot

### **Examples**

```
plotCCcomHeatmap(CPGscores)
```

```
plotCellProximityGenes
```

plotCellProximityGenes

## **Description**

Create visualization for cell proximity gene scores

#### Usage

```
plotCellProximityGenes(
  gobject,
  cpgObject,
  method = c("volcano", "cell_barplot", "cell-cell", "cell_sankey", "heatmap",
    "dotplot"),
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0.2,
  min_log2_fc = 0.5,
  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"
)
```

```
gobject giotto object
cpgObject cell proximity gene score object
method plotting method to use
min_cells minimum number of target cell type
min_int_cells minimum number of interacting cell type
```

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```
minimum adjusted p-value
min_fdr
                 minimum absolute spatial expression difference
min_spat_diff
min_log2_fc
                  minimum absolute log2 fold-change#' @param facet_scales ggplot facet scales
                  paramter
direction
                  differential expression directions to keep
cell_color_code
                  vector of colors with cell types as names
show_plot
                  show plots
return_plot
                  return plotting 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

plot

#### **Examples**

```
plotCPG(CPGscores)
```

plotCombineCCcom plotCombineCCcom

# Description

Create visualization for combined (pairwise) cell proximity gene scores

```
plotCombineCCcom(
  gobject,
  combCCcom,
  selected_LR = NULL,
  selected_cell_LR = NULL,
  detail_plot = T,
  simple_plot = F,
  simple_plot_facet = c("interaction", "genes"),
  facet_scales = "fixed",
  facet_ncol = length(selected_LR),
  facet_nrow = length(selected_cell_LR),
  colors = c("#9932CC", "#FF8C00"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCombineCCcom"
```

### **Arguments**

```
giotto object
gobject
combCCcom
                  combined communcation scores, output from combCCcom()
selected_LR
                  selected ligand-receptor pair
selected_cell_LR
                  selected cell-cell interaction pair for ligand-receptor pair
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
                  ggplot facet ncol parameter
facet_ncol
facet_nrow
                  ggplot facet nrow parameter
colors
                  vector with two colors to use
show_plot
                  show plots
return_plot
                  return plotting 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

ggplot

#### **Examples**

```
plotCombineCCcom(CPGscores)
```

```
plot {\tt Combine Cell Cell Communication} \\ plot {\tt Combine Cell Cell Communication} \\
```

# Description

Create visualization for combined (pairwise) cell proximity gene scores

```
plotCombineCellCellCommunication(
  gobject,
  combCCcom,
  selected_LR = NULL,
  selected_cell_LR = NULL,
  detail_plot = T,
  simple_plot = F,
```

```
simple_plot_facet = c("interaction", "genes"),
facet_scales = "fixed",
facet_ncol = length(selected_LR),
facet_nrow = length(selected_cell_LR),
colors = c("#9932CC", "#FF8C00"),
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "plotCombineCellCellCommunication")
```

# **Arguments**

```
gobject
                  giotto object
combCCcom
                  combined communcation scores, output from combCCcom()
                  selected ligand-receptor pair
selected_LR
selected_cell_LR
                  selected cell-cell interaction pair for ligand-receptor pair
detail_plot
                  show detailed info in both interacting cell types
simple_plot
                  show a simplified plot
simple_plot_facet
                  facet on interactions or genes with simple plot
facet_scales
                  ggplot facet scales paramter
facet_ncol
                  ggplot facet ncol parameter
facet_nrow
                  ggplot facet nrow parameter
colors
                  vector with two colors to use
show_plot
                  show plots
return_plot
                  return plotting 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

```
plotCombineCellCellCommunication(CPGscores)
```

```
plot {\tt Combine Cell Proximity Genes} \\ plot {\tt Combine Cell Proximity Genes}
```

### **Description**

Create visualization for combined (pairwise) cell proximity gene scores

### Usage

```
plotCombineCellProximityGenes(
  gobject,
  combCpgObject,
  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("#9932CC", "#FF8C00"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCombineCPG"
```

```
gobject
                  giotto object
combCpgObject
                  CPGscores, output from combineCellProximityGenes()
selected_interactions
                  interactions to show
selected_gene_to_gene
                  pairwise gene combinations to show
detail_plot
                  show detailed info in both interacting cell types
simple_plot
                  show a simplified plot
simple_plot_facet
                  facet on interactions or genes with simple plot
facet_scales
                  ggplot facet scales paramter
facet_ncol
                  ggplot facet ncol parameter
                  ggplot facet nrow parameter
facet_nrow
colors
                  vector with two colors to use
show_plot
                  show plots
return_plot
                  return plotting object
```

176 plotCombineCPG

#### Value

ggplot

### **Examples**

plotCombineCellProximityGenes(CPGscores)

plotCombineCPG

plotCombineCPG

### **Description**

Create visualization for combined (pairwise) cell proximity gene scores

### Usage

```
plotCombineCPG(
  gobject,
  combCpgObject,
  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("#9932CC", "#FF8C00"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCombineCPG"
)
```

plotCPG 177

```
show a simplified plot
simple_plot
simple_plot_facet
                  facet on interactions or genes with simple plot
                  ggplot facet scales paramter
facet_scales
                  ggplot facet ncol parameter
facet_ncol
facet_nrow
                  ggplot facet nrow parameter
colors
                  vector with two colors to use
show_plot
                  show plots
                  return plotting 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
```

#### Value

ggplot

### **Examples**

```
plotCombineCPG(CPGscores)
```

plotCPG plotCPG

# Description

Create visualization for cell proximity gene scores

```
plotCPG(
  gobject,
  cpgObject,
  method = c("volcano", "cell_barplot", "cell-cell", "cell_sankey", "heatmap",
    "dotplot"),
  min_cells = 5,
  min_int_cells = 3,
  min_fdr = 0.05,
  min_spat_diff = 0.2,
  min_log2_fc = 0.2,
  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"
)
```

178 plotCPGscores

### **Arguments**

gobject giotto object cpgObject cell proximity gene score object method plotting method to use min\_cells minimum number of target cell type min\_int\_cells minimum number of interacting cell type min\_fdr minimum adjusted p-value min\_spat\_diff minimum absolute spatial expression difference min\_log2\_fc minimum absolute log2 fold-change#' @param facet\_scales ggplot facet scales paramter differential expression directions to keep direction cell\_color\_code vector of colors with cell types as names show\_plot show plots return plotting object return\_plot 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

#### Value

plot

# Examples

```
plotCPG(CPGscores)
```

default\_save\_name

plotCPGscores

plotCPGscores

### **Description**

Create heatmap from cell-cell proximity scores

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

plotGTGscores 179

### **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
                  ggplot facet scales paramter
facet_scales
facet_ncol
                  ggplot facet ncol parameter
facet_nrow
                  ggplot facet nrow parameter
show_plot
                  show plot
```

#### **Details**

Give more details ...

#### Value

ggplot barplot

#### **Examples**

```
plotCPGscores(CPGscores)
```

plotGTGscores

plotGTGscores

### **Description**

Create heatmap from cell-cell proximity scores

```
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,
```

180 plotGTGscores

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

giotto object

# Arguments

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

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

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

show\_plot show plots

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

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

default\_save\_name

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

selected\_genes genes to show

#### **Details**

Give more details ...

# Value

ggplot barplot

```
plotGTGscores(GTGscore)
```

plotHeatmap 181

plotHeatmap

plotHeatmap

#### **Description**

Creates heatmap for genes and clusters.

## Usage

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

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

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

## **Details**

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

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

### Value

ggplot

## **Examples**

plotHeatmap(gobject)

plotly\_axis\_scale\_2D 183

```
plotly_axis_scale_2D plotly_axis_scale_2D
```

# Description

adjust the axis scale in 3D plotly plot

## Usage

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

#### **Arguments**

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

#### Value

edges in spatial grid as data.table()

## **Examples**

```
plotly_axis_scale_2D(gobject)
```

```
plotly_axis_scale_3D plotly_axis_scale_3D
```

## Description

adjust the axis scale in 3D plotly plot

## Usage

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

184 plotly\_grid

## **Arguments**

#### Value

edges in spatial grid as data.table()

## **Examples**

```
plotly_axis_scale_3D(gobject)
```

plotly\_grid

plotly\_grid

# Description

provide grid segment to draw in plot\_ly()

## Usage

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

# Arguments

```
spatial_grid spatial_grid in giotto object
```

## Value

edges in spatial grid as data.table()

## **Examples**

```
plotly_grid(gobject)
```

plotly\_network 185

plotly\_network

plotly\_network

## **Description**

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

## Usage

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

#### **Arguments**

gobject network in giotto object

#### Value

edges in network as data.table()

### **Examples**

```
plotly_network(gobject)
```

```
plotMetaDataCellsHeatmap
```

plotMetaDataCellsHeatmap

# 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"
    )
Arguments
    gobject
                     giotto object
                     annotation columns found in pDataDT(gobject)
    metadata_cols
    spat_enr_names spatial enrichment results to include
    value_cols
                     value columns to use
    first_meta_col if more than 1 metadata column, select the x-axis factor
    second_meta_col
                     if more than 1 metadata column, select the facetting factor
    show_values
                     which values to show on heatmap
    custom_cluster_order
                     custom cluster order (default = NULL)
    clus_cor_method
                     correlation method for clusters
    clus\_cluster\_method
                     hierarchical cluster method for the clusters
                     midpoint of show_values
    midpoint
                     size of x-axis text
    x_text_size
    x_text_angle
                     angle of x-axis text
    y_text_size
                     size of y-axis text
    strip_text_size
                     size of strip text
    show_plot
                     show plot
```

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

directly save the plot [boolean]

return ggplot object

return\_plot

save\_plot

plotMetaDataHeatmap 187

#### **Details**

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

## Value

ggplot or data.table

#### See Also

plotMetaDataHeatmap for gene expression instead of numeric cell annotation data.

#### **Examples**

```
plotMetaDataCellsHeatmap(gobject)
```

```
plotMetaDataHeatmap \hspace{3cm} plotMetaDataHeatmap
```

## **Description**

Creates heatmap for genes within aggregated clusters.

## Usage

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

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

## Arguments

```
gobject
                  giotto object
expression_values
                  expression values to use
                 annotation columns found in pDataDT(gobject)
metadata_cols
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)
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 of show_values
midpoint
                  size of x-axis text
x_text_size
x_text_angle
                  angle of x-axis text
                  size of y-axis text
y_text_size
strip_text_size
                  size of strip text
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name
```

#### **Details**

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

plotPCA 189

#### Value

ggplot or data.table

#### See Also

plotMetaDataCellsHeatmap for numeric cell annotation instead of gene expression.

#### **Examples**

```
plotMetaDataHeatmap(gobject)
```

plotPCA

plotPCA

#### **Description**

Short wrapper for PCA visualization

giotto object

## Usage

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

## **Arguments**

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)
                 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
```

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```
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
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
show_legend
                  show legend
                  title for plot, defaults to cell_color parameter
title
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
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
```

plotPCA\_2D 191

#### **Details**

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

#### Value

ggplot

## **Examples**

```
plotPCA(gobject)
```

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
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)
                 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
```

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```
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
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
                  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
```

plotPCA\_3D 193

#### **Details**

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

#### Value

ggplot

#### **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
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
select_cell_groups
                 select subset of cells/clusters based on cell_color parameter
```

194 plotRankSpatvsExpr

```
select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
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
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
```

# **Details**

Description of parameters.

### Value

plotly

## **Examples**

plotPCA\_3D(gobject)

 $plotRankSpatvsExpr \\ plotRankSpatvsExpr$ 

## Description

Plots dotplot to compare ligand-receptor rankings from spatial and expression information

plotRankSpatvsExpr 195

### Usage

```
plotRankSpatvsExpr(
  gobject,
  combCC,
  expr_rnk_column = "LR_expr_rnk",
  spat_rnk_column = "LR_spat_rnk",
  midpoint = 10,
  size\_range = c(0.01, 1.5),
  xlims = NULL,
  ylims = NULL,
  selected_ranks = c(1, 10, 20),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotRankSpatvsExpr"
)
```

## **Arguments**

gobject giotto object combCC combined communinication scores from combCCcom expr\_rnk\_column column with expression rank information to use spat\_rnk\_column column with spatial rank information to use midpoint midpoint of colors size ranges of dotplot size\_range xlims x-limits, numerical vector of 2 vlims y-limits, numerical vector of 2 selected\_ranks numerical vector, will be used to print out the percentage of top spatial ranks are recovered show\_plot show plots return\_plot return plotting 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**

```
plotRankSpatvsExpr(CPGscores)
```

196 plotRecovery

plotRecovery plotRecovery

#### **Description**

Plots recovery plot to compare ligand-receptor rankings from spatial and expression information

## Usage

```
plotRecovery(
  gobject,
  combCC,
  expr_rnk_column = "exprPI_rnk",
  spat_rnk_column = "spatPI_rnk",
  ground_truth = c("spatial", "expression"),
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotRecovery"
)
```

## **Arguments**

```
gobject
                  giotto object
combCC
                  combined communinication scores from combCCcom
expr_rnk_column
                  column with expression rank information to use
spat_rnk_column
                  column with spatial rank information to use
ground_truth
                  what to consider as ground truth (default: spatial)
                  show plots
show_plot
return_plot
                  return plotting 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

ggplot

## **Examples**

```
plotRecovery(CPGscores)
```

plotRecovery\_sub 197

```
plotRecovery_sub plotRecovery_sub
```

### **Description**

Plots recovery plot to compare ligand-receptor rankings from spatial and expression information

## Usage

```
plotRecovery_sub(combCC, first_col = "LR_expr_rnk", second_col = "LR_spat_rnk")
```

## **Arguments**

```
combCC combined communinication scores from combCCcom
first_col first column to use
```

second\_col second column to use

## **Examples**

```
plotRecovery_sub(CPGscores)
```

```
plotStatDelaunayNetwork
```

plot Stat Delaunay Network

## **Description**

Plots network statistics for a Delaunay network..

# Usage

```
plotStatDelaunayNetwork(
  gobject,
  dimensions = c("sdimx", "sdimy"),
  name = "delaunay_network",
  maximum_distance = "auto",
  minimum_k = 0,
  Y = TRUE,
  j = TRUE,
  S = 0,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotStatDelaunayNetwork",
  ...
)
```

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### **Arguments**

gobject giotto object

dimensions which spatial dimensions to use (maximum 2 dimensions)

name name for spatial network (default = 'delaunay\_network')

maximum\_distance

distance cuttof for Delaunay neighbors to consider

minimum\_k minimum neighbours if maximum\_distance != NULL

Y (RTriangle) If TRUE prohibits the insertion of Steiner points on the mesh bound-

ary.

j (RTriangle) If TRUE jettisons vertices that are not part of the final triangulation

from the output.

S (RTriangle) Specifies the maximum number of added Steiner points.

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

... Other parameters of the triangulate function

### Details

Plots statistics for a spatial Delaunay network as explained in triangulate. This can be used to further finetune the createDelaunayNetwork function.

#### Value

giotto object with updated spatial network slot

## **Examples**

 $\verb|plotStatDelaunayNetwork(gobject)|$ 

plotTSNE plotTSNE

## Description

Short wrapper for tSNE visualization

## Usage

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

plotTSNE 199

### **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 subset of cells based on cell IDs select\_cells show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges

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point\_size size of point (cell)

point\_border\_col

color of border around points

 ${\tt point\_border\_stroke}$ 

stroke size of border around points

title title for plot, defaults to cell\_color parameter

show\_legend show legend

 ${\tt legend\_symbol\_size}$ 

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

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 plotTSNE\_3D

## Value

ggplot

# **Examples**

plotTSNE(gobject)

plotTSNE\_2D plotTSNE\_2D

## **Description**

Short wrapper for tSNE visualization

plotTSNE\_2D 201

#### Usage

```
plotTSNE_2D(
     gobject,
     dim_reduction_name = "tsne",
     default_save_name = "tSNE_2D",
Arguments
   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 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

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

other\_point\_size

size of not selected cells

show\_cluster\_center

plot center of selected clusters

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```
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

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

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

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

#### **Details**

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

### Value

ggplot

## **Examples**

```
plotTSNE_2D(gobject)
```

plotTSNE\_3D 203

plotTSNE\_3D plotTSNE\_3D

## **Description**

Visualize cells according to dimension reduction coordinates

## Usage

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

```
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
```

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

#### **Details**

Description of parameters.

#### Value

plotly

# **Examples**

```
plotTSNE_3D(gobject)
```

plotUMAP plotUMAP

## **Description**

Short wrapper for UMAP visualization

## Usage

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

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 subset of cells based on cell IDs select\_cells show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point 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 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

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

```
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]

# Details

save\_param

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",
   ...
)
```

plotUMAP\_2D 207

### **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 subset of cells based on cell IDs select\_cells show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges

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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

 ${\tt legend\_symbol\_size}$ 

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

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\_2D(gobject)

plotUMAP\_3D plotUMAP\_3D

## Description

Visualize cells according to dimension reduction coordinates

plotUMAP\_3D 209

#### Usage

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

```
gobject
                  giotto object
dim_reduction_name
                  umap dimension reduction name
default_save_name
                  default save name for saving, don't change, change save_name in save_param
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
                  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)
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
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]

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
```

```
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
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
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
gobject
                  giotto object
```

## **Details**

Description of parameters.

## Value

ggplot

## **Examples**

```
plot_point_layer_ggplot(gobject)
```

## **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
```

```
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
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_legend
                  show legend
                  giotto object
gobject
```

### **Details**

Description of parameters.

### Value

ggplot

## **Examples**

```
plot_point_layer_ggplot_noFILL(gobject)
```

## **Description**

creat ggplot point layer for spatial coordinates

#### Usage

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

```
x-axis dimension name (default = 'sdimx')
sdimx
                  y-axis dimension name (default = 'sdimy')
sdimv
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
{\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 subset of cells based on cell IDs
select_cells
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
show_other_cells
                  display not selected cells
other_cell_color
                  color for not selected cells
other_point_size
                  point size for not selected cells
show_legend
                  show legend
                  giotto object
gobject
```

#### **Details**

Description of parameters.

## Value

ggplot

#### **Examples**

```
plot_spat_point_layer_ggplot(gobject)
```

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

## Description

creat ggplot point layer for spatial coordinates without borders

#### Usage

```
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,
  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

```
sdimx
                  x-axis dimension name (default = 'sdimx')
sdimy
                  y-axis dimension name (default = 'sdimy')
{\tt 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
                  size of point (cell)
point_size
```

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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
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
gobject
                  giotto object
```

#### **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 219

rankEnrich

rankEnrich

#### **Description**

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

### Usage

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

#### **Arguments**

### **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

```
make Sign Matrix Rank
```

220 rankSpatialCorGroups

#### **Examples**

```
rankEnrich(gobject)
```

rankPermutation rankPermutation

# **Description**

creates permutation for the rankEnrich test

#### Usage

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

# **Examples**

```
rankPermutation()
```

rankSpatialCorGroups rankSpatialCorGroups

# **Description**

Rank spatial correlated clusters according to correlation structure

# Usage

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

# **Arguments**

```
gobject giotto object

spatCorObject spatial correlation object

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

show_plot show plot

return_plot return ggplot object

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

rank\_binarize 221

# Value

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

# **Examples**

```
rankSpatialCorGroups(gobject)
```

rank\_binarize

rank\_binarize

# Description

create binarized scores using arbitrary rank of top genes

# Usage

```
rank_binarize(x, max_rank = 200)
```

readGiottoInstructions

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()

222 removeGeneAnnotation

remove Cell Annotation remove Cell Annotation

# Description

removes cell annotation of giotto object

# Usage

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

#### **Arguments**

gobject giotto object

columns names of columns to remove

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

#### **Details**

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

# Value

giotto object

# **Examples**

removeCellAnnotation(gobject)

removeGeneAnnotation removeGeneAnnotation

# **Description**

removes gene annotation of giotto object

# Usage

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

# **Arguments**

gobject giotto object

columns names of columns to remove

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

# **Details**

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

#### Value

giotto object

# **Examples**

```
removeGeneAnnotation(gobject)
```

```
replaceGiottoInstructions
```

replace Giot to Instructions

# **Description**

Function to replace all instructions from giotto object

# Usage

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

# **Arguments**

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

## Value

named vector with giotto instructions

# **Examples**

```
replaceGiottoInstructions()
```

runPCA

runPCA

# Description

runs a Principal Component Analysis

# 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,
  ...
)
```

224 runtSNE

### **Arguments**

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

#### **Details**

See PCA for more information about other parameters.

# Value

giotto object with updated PCA dimension recuction

# **Examples**

```
runPCA(gobject)
```

runtSNE

runtSNE

# Description

run tSNE

# 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,
)
```

runtSNE 225

# **Arguments**

gobject giotto object

expression\_values

expression values to use

reduction cells or genes

dim\_reduction\_to\_use

use another dimension reduction set as input

dim\_reduction\_name

name of dimension reduction set to use

dimensions\_to\_use

number of dimensions to use as input

name arbitrary name for tSNE run

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

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

dims tSNE param: number of dimensions to return

perplexity tSNE param: perplexity

theta tSNE param: theta

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

set\_seed use of seed

seed\_number seed number to use

... additional tSNE parameters

#### **Details**

See Rtsne for more information about these and other parameters.

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

#### Value

giotto object with updated tSNE dimension recuction

# **Examples**

runtSNE(gobject)

226 runUMAP

runUMAP

runUMAP

# **Description**

run UMAP

# Usage

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

# **Arguments**

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

selectPatternGenes 227

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

#### **Details**

See umap for more information about these and other parameters.

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

#### Value

giotto object with updated UMAP dimension recuction

#### **Examples**

```
runUMAP(gobject)
```

selectPatternGenes selectPatternGenes

### **Description**

Select genes correlated with spatial patterns

#### Usage

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

#### **Arguments**

spatPatObj

```
dimensions dimensions to identify correlated genes for.

top_pos_genes Top positively correlated genes.

top_neg_genes Top negatively correlated genes.

min_pos_cor Minimum positive correlation score to include a gene.

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

Output from detectSpatialPatterns

228 show,giotto-method

#### **Details**

Description.

#### Value

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

# **Examples**

```
selectPatternGenes(gobject)
```

```
select_expression_values
```

select\_expression\_values

# Description

helper function to select expression values

# Usage

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

# Arguments

gobject giotto object

values expression values to extract

# Value

expression matrix

show,giotto-method

show method for giotto class

# Description

show method for giotto class

# Usage

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

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

# **Description**

Creates dendrogram for selected clusters.

#### Usage

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

### **Arguments**

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

# **Details**

Expression correlation dendrogram for selected clusters.

230 showClusterHeatmap

#### Value

ggplot

# **Examples**

showClusterDendrogram(gobject)

showClusterHeatmap

*showClusterHeatmap* 

### **Description**

Creates heatmap based on identified clusters

# Usage

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

# Arguments

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

showCPGscores 231

#### **Details**

Correlation heatmap of selected clusters.

#### Value

ggplot

# **Examples**

showClusterHeatmap(gobject)

showCPGscores

showCPGscores

# **Description**

visualize Cell Proximity Gene enrichment scores

### Usage

```
showCPGscores(
 gobject,
 CPGscore,
 method = c("volcano", "cell_barplot", "cell-cell", "cell_sankey", "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
min_spat_diff spatial difference threshold
min_log2_fc min log2 fold-change
keep_int_duplicates
keep both cell_A-cell_B and cell_B-cell_A
```

#### **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 ...

showGiottoInstructions 233

#### Value

ggplot barplot

# **Examples**

```
showGeneExpressionProximityScore(scores)
```

```
showGiottoInstructions
```

showGiottoInstructions

# **Description**

Function to display all instructions from giotto object

# Usage

```
showGiottoInstructions(gobject)
```

# **Arguments**

gobject giotto object

#### Value

named vector with giotto instructions

# **Examples**

showGiottoInstructions()

showGTGscores

showGTGscores

# Description

visualize Cell Proximity Gene enrichment scores

# Usage

```
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,
```

showGTGscores

```
specific_genes_1 = NULL,
specific_genes_2 = NULL,
first_cell_name = "ligand cell",
second_cell_name = "receptor cell",
return_DT = F
```

# Arguments

method visualization method min\_cells min number of cells threshold min\_pval p-value threshold spatial difference threshold min\_spat\_diff min\_log2\_fc log2 fold-change threshold 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()

#### **Details**

Give more details ...

**CPGscore** 

#### 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

# Usage

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

# **Arguments**

# **Details**

Give more details ...

## Value

ggplot barplot

# **Examples**

showIntExpressionProximityScore(scores)

236 showPattern

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

# Value

ggplot

## See Also

showPattern2D

# **Examples**

```
showPattern(gobject)
```

showPattern2D 237

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
                 giotto object
                 Output from detectSpatialPatterns
spatPatObj
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 of ggplot
show_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 for saving, don't change, change save\_name in save\_param

#### Value

ggplot

# **Examples**

```
{\tt showPattern2D(gobject)}
```

238 showPattern3D

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 ends of the PC values.
trim
background_color
                  background color for plot
grid_border_color
                  color for grid
                  show legend of plot
show_legend
point_size
                  adjust the point size
axis_scale
                  scale the axis
                  cutomize the scale of the axis
custom_ratio
x_ticks
                  the tick number of x_axis
                  the tick number of y_axis
y_ticks
z_ticks
                  the tick number of z_axis
```

showPatternGenes 239

```
show_plot show plot

return_plot return plot 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

plotly

# **Examples**

```
showPattern3D(gobject)
```

showPatternGenes

showPatternGenes

# **Description**

show genes correlated with spatial patterns

# Usage

```
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"
)
```

# **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

240 showProcessingSteps

```
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

show Processing Steps

# 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 241

showSpatialCorGenes showSpatialCorGenes

# Description

Shows and filters spatially correlated genes

# Usage

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

# **Arguments**

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

# Value

data.table with filtered information

# **Examples**

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

242 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 243

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

#### **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)
```

```
sort_combine_two_DT_columns
sort_combine_two_DT_columns
```

# **Description**

fast sorting and pasting of 2 character columns

#### Usage

```
sort_combine_two_DT_columns(DT, column1, column2, myname = "unif_gene_gene")
```

# **Examples**

```
sort_combine_two_DT_columns()
```

spatCellCellcom 245

spatCellCellcom spatCellCellcom

### **Description**

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

### Usage

```
spatCellCellcom(
  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,
 adjust_method = c("fdr", "bonferroni", "BH", "holm", "hochberg", "hommel", "BY",
    "none"),
  adjust_target = c("genes", "cells"),
  do_parallel = TRUE,
 cores = NA,
  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
                  number of iterations
random_iter
                  first specific gene set from gene pairs
gene_set_1
gene_set_2
                  second specific gene set from gene pairs
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
                  which method to adjust p-values
adjust_method
                  adjust multiple hypotheses at the cell or gene level
adjust_target
                  run calculations in parallel with mclapply
do_parallel
                  number of cores to use if do_parallel = TRUE
cores
                  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.

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#### Value

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

#### **Examples**

```
spatCellCellcom(gobject)
```

spatCellPlot

spatCellPlot

# **Description**

Visualize cells according to spatial coordinates

# Usage

```
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,
  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,
```

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```
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
    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_cells
                     select subset of cells based on cell IDs
    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
    label_size
                     size of labels
    label_fontface font of labels
```

show underlying spatial network

show\_network

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```
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

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

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

# **Examples**

spatCellPlot(gobject)

spatCellPlot2D 249

spatCellPlot2D

spatCellPlot2D

# **Description**

Visualize cells according to spatial coordinates

# Usage

```
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",
 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,
```

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```
cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatCellPlot2D"
Arguments
    gobject
                      giotto object
    sdimx
                      x-axis dimension name (default = 'sdimx')
    sdimy
                      y-axis dimension name (default = 'sdimy')
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                      numeric cell annotation columns
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
                      select 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
    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

show\_grid

alpha of spatial network

show spatial grid

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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

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

show\_plot show plot

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

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

default\_save\_name

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

# **Details**

Description of parameters.

# Value

ggplot

# **Examples**

spatCellPlot2D(gobject)

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spatDimCellPlot

spatDimCellPlot

#### **Description**

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

### Usage

```
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} \boldsymbol{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 256 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**

```
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
```

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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)
```

### **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 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
    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

#### **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,
```

```
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
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
```

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

#### **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,
```

```
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**

)

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

```
axis_title size of axis title

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

### **Details**

Description of parameters.

#### Value

ggplot

### See Also

```
spatDimPlot3D
```

### **Examples**

```
spatDimPlot2D(gobject)
```

spatDimPlot3D

spatDimPlot3D

### **Description**

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

```
spatDimPlot3D(
 gobject,
 plot_alignment = c("horizontal", "vertical"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2_to_use = 2,
 dim3_to_use = 3,
 sdimx = "sdimx",
  sdimy = "sdimy",
 sdimz = "sdimz",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
```

```
label_size = 16,
     select_cell_groups = NULL,
     select_cells = NULL,
     show_other_cells = T,
     other_cell_color = "lightgrey",
     other_point_size = 1.5,
     cell_color = NULL,
     color_as_factor = T,
     cell_color_code = NULL,
     dim_point_size = 3,
     nn_network_alpha = 0.5,
     show_spatial_network = F,
     spatial_network_name = "spatial_network",
     network_color = "lightgray",
     spatial_network_alpha = 0.5,
     show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
     spatial_grid_color = NULL,
     spatial_grid_alpha = 0.5,
     spatial_point_size = 3,
     axis_scale = c("cube", "real", "custom"),
     custom_ratio = NULL,
     x_ticks = NULL,
     y_ticks = NULL,
     z_ticks = NULL,
     legend_text_size = 12,
     show_plot = NA,
     return_plot = NA,
     save_plot = NA,
     save_param = list(),
     default_save_name = "spatDimPlot3D"
Arguments
   gobject
                   giotto object
   plot_alignment direction to align plot
   dim_reduction_to_use
                   dimension reduction to use
   dim_reduction_name
                   dimension reduction name
   dim1_to_use
                   dimension to use on x-axis
   dim2_to_use
                   dimension to use on y-axis
```

sdimx = spatial dimension to use on x-axis

sdimy = spatial dimension to use on y-axis

sdimz = spatial dimension to use on z-axis

show\_NN\_network

show underlying NN network

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

dimension to use on z-axis

dim3\_to\_use

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
```

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```
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
                  cowplot param: relative height
cow_rel_h
                  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

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

# **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 283

```
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)

284 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 285

```
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

286 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 287

### 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)
```

### **Examples**

```
spatialDE(gobject)
```

288 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

### **Examples**

```
Spatial_AEH(gobject)
```

Spatial\_DE 289

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)
```

 ${\tt spatNetwDistributions} \textit{DistributionsDistance}$ 

# Description

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

290 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 293

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 295

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.

296 spatPlot2D

### 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 297

```
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 cow\_rel\_w

cowplot param: how to align

show plot

cow\_align

show\_plot

spatPlot2D\_single 299

### **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 302 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,
```

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```
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 of plot
title
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
                  set the number of ticks on the y-axis
y_ticks
z_ticks
                  set the number of ticks on the z-axis
                  show plot
show_plot
                  return 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

304 spat\_OR\_func

## **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, calc_hub = F, hub_min_int = 3)
```

spat\_OR\_func

 $spat\_OR\_func$ 

# Description

calculate odds-ratio

```
spat_OR_func(gene, bin_matrix, spat_mat, calc_hub = F, hub_min_int = 3)
```

```
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,
 adjust_method = c("fdr", "bonferroni", "BH", "holm", "hochberg", "hommel", "BY",
    "none"),
  adjust_target = c("genes", "cells"),
  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
cell_type_1
                  first cell type
cell_type_2
                  second cell type
gene_set_1
                  first specific gene set from gene pairs
                  second specific gene set from gene pairs
gene_set_2
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
                  which method to adjust p-values
adjust_method
adjust_target
                  adjust multiple hypotheses at the cell or gene level
verbose
                  verbose
```

306 stitchFieldCoordinates

### **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**

```
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**

```
split_dendrogram_in_two(dend)
```

stitchFieldCoordinates

stitchFieldCoordinates

# Description

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

stitchFieldCoordinates 307

#### Usage

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

## **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

```
stitchFieldCoordinates(gobject)
```

308 subClusterCells

```
stitchTileCoordinates stitchTileCoordinates
```

# Description

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

## Usage

```
stitchTileCoordinates(location_file, Xtilespan, Ytilespan)
```

### **Arguments**

```
\begin{array}{ll} \mbox{location\_file} & \mbox{location dataframe with $X$ and $Y$ coordinates} \\ \mbox{Xtilespan} & \mbox{numerical value specifying the width of each tile} \\ \mbox{Ytilespan} & \mbox{numerical value specifying the height of each tile} \\ \end{array}
```

#### **Details**

•••

### **Examples**

```
stitchTileCoordinates(gobject)
```

subClusterCells

subClusterCells

## **Description**

subcluster cells

```
subClusterCells(
 gobject,
 name = "sub_clus",
 cluster_method = c("leiden", "louvain_community", "louvain_multinet"),
 cluster_column = NULL,
  selected_clusters = NULL,
 hvg_param = list(reverse_log_scale = T, difference_in_variance = 1, expression_values
   = "normalized"),
 hvg_min_perc_cells = 5,
 hvg_mean_expr_det = 1,
 use_all_genes_as_hvg = FALSE,
 min_nr_of_hvg = 5,
 pca_param = list(expression_values = "normalized", scale_unit = T),
 nn_param = list(dimensions_to_use = 1:20),
 k_neighbors = 10,
 resolution = 1,
```

subClusterCells 309

```
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 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 parameters for parameters for createNearestNetwork nn\_param k\_neighbors number of k for createNearestNetwork resolution resolution gamma gamma omega omega specify specific path to python if required python\_path nn\_network\_to\_use type of NN network to use (kNN vs sNN) name of NN network to use network\_name return\_gobject boolean: return giotto object (default = TRUE)

#### **Details**

verbose

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

verbose

• 5. do clustering

310 subsetGiotto

### Value

giotto object with new subclusters appended to cell metadata

# See Also

 ${\tt doLouvainCluster\_multinet}, {\tt doLouvainCluster\_community} \ and \ @see also \ {\tt doLeidenCluster}$ 

# **Examples**

```
subClusterCells(gobject)
```

subsetGiotto

subsetGiot to

# Description

subsets Giotto object including previous analyses.

# Usage

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

# Arguments

gobject giotto object
cell\_ids cell IDs to keep
gene\_ids gene IDs to keep
verbose be verbose

# Value

giotto object

# **Examples**

subsetGiotto(gobject)

subsetGiottoLocs 311

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)
```

312 trendSceek

trendSceek	trendSceek
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## **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)
```

viewHMRFresults 313

viewHMRFresults

viewHMRFresults

# Description

View results from doHMRF.

# Usage

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

# **Arguments**

```
gobject giotto object

HMRFoutput HMRF output from doHMRF

k number of HMRF domains

betas_to_view results from different betas that you want to view

... paramters to visPlot()
```

# **Details**

Description ...

# Value

spatial plots with HMRF domains

## See Also

```
visPlot
```

```
viewHMRFresults(gobject)
```

314 viewHMRFresults2D

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

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)
```

viewHMRFresults3D 315

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)
```

316 violinPlot

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
```

visDimGenePlot 317

### 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
)
```

318 visDimGenePlot

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

show\_legend show legend show\_plots show plots

## **Details**

Description of parameters.

#### Value

ggplot

# Examples

visDimGenePlot(gobject)

## **Description**

Visualize cells and gene expression according to dimension reduction coordinates

## Usage

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

## **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.

322 visDimPlot

### 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
    {\tt 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

### **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

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

```
visDimPlot_2D_ggplot(gobject)
```

```
visDimPlot_2D_plotly visDimPlot_2D_plotly
```

## **Description**

Visualize cells according to dimension reduction coordinates

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

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

## **Arguments**

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

#### **Details**

Description of parameters.

#### Value

plotly

## **Examples**

```
visDimPlot_2D_plotly(gobject)
```

```
visDimPlot_3D_plotly
```

## **Description**

Visualize cells according to dimension reduction coordinates

#### Usage

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

# **Arguments**

```
giotto object
gobject
dim_reduction_to_use
                 dimension reduction to use
dim_reduction_name
                 dimension reduction name
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
dim3_to_use
                 dimension to use on z-axis
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_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
dim2_to_use
                  dimension to use on y-axis
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
```

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#### **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 genes to show

genes\_high\_color

color represents high gene expression

genes\_mid\_color

color represents middle gene expression

genes\_low\_color

color represents low gene expression

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

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

spatial\_grid\_name

name of spatial grid to use

midpoint expression midpoint
scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

show\_legend show legend

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

# Details

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
```

```
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

network\_color

show\_grid

spatial\_network\_name

genes\_high\_color

genes\_mid\_color

genes\_low\_color

spatial\_grid\_name

point\_size show\_legend

axis\_scale

x\_ticks

y\_ticks

#### Usage

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

color of spatial network

show spatial grid

name of spatial network to use

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

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```
z_ticks number of ticks on z axis
show_plots show plots
grid_color color of spatial grid
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
```

#### **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
```

color of not selected cells

color of spatial network

show underlying spatial network

name of spatial network to use

other\_cell\_color

show\_network

network\_color

spatial\_network\_name

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```
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,
```

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```
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
```

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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**

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

visPlot\_2D\_plotly 341

```
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
```

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```
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,
```

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```
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
```

#### **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,
```

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

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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 for the label label\_size

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

three modes to adjust axis scale ratio axis\_scale 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

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

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

show\_legend show legend show\_plot show plot

#### **Details**

Description of parameters.

#### Value

ggplot or plotly

## **Examples**

visSpatDimGenePlot(gobject)

```
visSpatDimGenePlot_2D visSpatDimGenePlot_2D
```

## **Description**

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

## Usage

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

#### **Arguments**

```
gobject giotto object
```

expression\_values

gene expression values to use

plot\_alignment direction to align plot

genes genes to show

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimension reduction name

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

point\_size size of point (cell)

dim\_point\_border\_col

color of border around points

dim\_point\_border\_stroke

stroke size of border around points

show\_NN\_network

show underlying NN network

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

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

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

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

spatial\_network\_name

name of spatial network to use

spatial\_grid\_name

name of spatial grid to use

spatial\_point\_size

spatial plot: point size

spatial\_point\_border\_col

color of border around points

 ${\tt spatial\_point\_border\_stroke}$ 

stroke size of border around points

midpoint size of point (cell)

cow\_n\_col cowplot param: how many columns cow\_rel\_h cowplot param: relative height

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

show\_legend show legend

dim\_point\_size dim reduction plot: point size

show\_plot show plot

#### **Details**

Description of parameters.

#### Value

ggplot

#### **Examples**

```
visSpatDimGenePlot_2D(gobject)
```

```
visSpatDimGenePlot_3D visSpatDimGenePlot_3D
```

## **Description**

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

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

y\_ticks = NULL, z\_ticks = NULL

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

## **Details**

Description of parameters.

show legend show plot

show\_legend

show\_plot

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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,
```

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 $show_NN_network = F,$ 

network\_name
cell\_color

```
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
```

#### **Details**

Description of parameters.

#### Value

ggplot

save\_name
save\_format

show\_saved\_plot

## **Examples**

```
visSpatDimPlot_2D(gobject)
```

name of plot

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

load & display the saved plot

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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

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

# **Examples**

```
visSpatDimPlot_3D(gobject)
```

writeHMRFresults

writeHMRFresults

#### **Description**

write results from doHMRF to a data.table.

#### Usage

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

## Arguments

gobject giotto object

HMRF output HMRF output from doHMRF

k k to write results for

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

print\_command see the python command

## Value

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

## **Examples**

```
writeHMRFresults(gobject)
```

## **Description**

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

```
write_giotto_viewer_annotation(
  annotation,
  annot_name = "test",
  output_directory = getwd()
)
```

#### **Arguments**

```
annotation annotation from the data.table from giotto object
annot_name name of the annotation
output_directory
directory where to save the files
```

#### Value

write a .txt and .annot file for the selection annotation

#### **Description**

write out dimensional reduction data from a giotto object for the Viewer

## Usage

```
write_giotto_viewer_dim_reduction(
  dim_reduction_cell,
  dim_red = NULL,
  dim_red_name = NULL,
  dim_red_rounding = NULL,
  dim_red_rescale = c(-20, 20),
  output_directory = getwd()
)
```

## **Arguments**

```
dim_reduction_cell

dimension reduction slot from giotto object

dim_red high level name of dimension reduction

dim_red_name specific name of dimension reduction to use

dim_red_rounding

numerical indicating how to round the coordinates

dim_red_rescale

numericals to rescale the coordinates

output_directory

directory where to save the files
```

#### Value

write a .txt and .annot file for the selection annotation

```
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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