# Package 'Giotto'

May 29, 2020

```
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Version 0.3.5
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BugReports https://github.com/RubD/Giotto/issues
RoxygenNote 7.1.0
Depends data.table (>= 1.12.2),
      ggplot2 (>= 3.1.1),
      base (>= 3.5.1),
      utils (>= 3.5.1),
      R (>= 3.5.1)
Imports Matrix,
      magick,
      matrixStats (\geq 0.55.0),
      uwot (>= 0.0.0.9010),
      cowplot (>= 0.9.4),
      grDevices,
      RColorBrewer (>= 1.1-2),
      dbscan (>= 1.1-3),
      ggalluvial (>= 0.9.1),
      scales (>= 1.0.0),
      ComplexHeatmap (>= 1.20.0),
      \frac{1}{2} qvalue (>= 2.14.1),
      If a (>= 1.12.0),
      igraph (>= 1.2.4.1),
      irlba,
      plotly,
     parallel,
      reticulate (>= 1.14),
      magrittr,
      limma,
      ggdendro,
```

smfishHmrf,

2 R topics documented:

```
devtools,
     reshape2,
      ggraph,
      Rcpp,
      Rfast,
      Rtsne (>= 0.15),
      rlang (>= 0.4.3),
      R.utils,
      fitdistrplus,
      deldir
Suggests knitr,
     rmarkdown,
     MAST,
      scran (>= 1.10.1),
     png,
     FactoMineR,
     tiff,
     biomaRt,
     trendsceek,
      multinet (>= 3.0.2),
      RTriangle (>= 1.6-0.10)
biocViews
VignetteBuilder knitr
LinkingTo Rcpp,
      RcppArmadillo
Remotes lambdamoses/smfishhmrf-r
```

# R topics documented:

adapt_aspect_ratio
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addCellMetadata
addCellStatistics
addGeneMetadata
addGenesPerc
addGeneStatistics
addGiottoImage
addGiottoImageToSpatPlot
addHMRF
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adapt\_aspect\_ratio
adapt\_aspect\_ratio

#### **Description**

adapt the aspact ratio after inserting cross section mesh grid lines

#### Usage

```
adapt_aspect_ratio(
  current_ratio,
  cell_locations,
  sdimx = NULL,
  sdimy = NULL,
  sdimz = NULL,
  mesh_obj = NULL
)
```

addCellIntMetadata

addCellIntMetadata

#### **Description**

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

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

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#### **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

addCellMetadata

## **Description**

adds cell metadata to the giotto object

#### Usage

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

# Arguments

gobject giotto object

new\_metadata new cell metadata to use (data.table, data.frame, ...)
vector\_name (optional) custom name if you provide a single vector

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

addCellStatistics 11

#### **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

## 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

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#### 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

## **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**

 ${\it addGeneMetadata(gobject)}\\$ 

addGenesPerc 13

addGenesPerc

addGenesPerc

## **Description**

calculates the total percentage of (normalized) counts for a subset of selected genes

#### Usage

```
addGenesPerc(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  vector_name = "gene_perc",
  return_gobject = TRUE
)
```

## **Arguments**

```
gobject giotto object
expression_values
expression values to use
genes vector of selected genes
vector_name column name as seen in pDataDT()
return_gobject boolean: return giotto object (default = TRUE)
```

## Value

giotto object if return\_gobject = TRUE, else a vector with

# **Examples**

```
addGenesPerc(gobject)
```

 ${\tt addGeneStatistics}$ 

addGeneStatistics

## **Description**

adds gene statistics to the giotto object

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

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#### **Arguments**

## **Details**

This function will add the following statistics to gene metadata:

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

#### Value

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

## **Examples**

```
addGeneStatistics(gobject)
```

 ${\tt addGiottoImage}$ 

addGiottoImage

# Description

Adds giotto image objects to your giotto object

# Usage

```
addGiottoImage(gobject, images)
```

# **Arguments**

gobject giotto object

images list of giotto image objects, see createGiottoImage

#### Value

an updated Giotto object with access to the list of images

```
addGiottoImage(mg_object)
```

addGiottoImageToSpatPlot

add Giot to Image To Spat Plot

## **Description**

Add a giotto image to a spatial ggplot object post creation

# Usage

```
addGiottoImageToSpatPlot(spatpl = NULL, gimage = NULL)
```

## **Arguments**

spatpl a spatial ggplot object

gimage a giotto image, see createGiottoImage

## Value

an updated spatial ggplot object

# **Examples**

addGiottoImageToSpatPlot(mg\_object)

addHMRF

addHMRF

# Description

Add selected results from doHMRF to the giotto object

# Usage

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

## Arguments

gobject giotto object

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

k number of domains

name specify a custom name

# Details

Description ...

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#### Value

```
giotto object
```

#### **Examples**

```
addHMRF(gobject)
```

addNetworkLayout

addNetworkLayout

## **Description**

Add a network layout for a selected nearest neighbor network

# Usage

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

## **Arguments**

#### Details

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

## Value

giotto object with updated layout for selected NN network

```
addNetworkLayout(gobject)
```

addStatistics 17

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

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#### 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

```
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)
```

```
annotateSpatialNetwork
```

annotate Spatial Network

## **Description**

Annotate spatial network with cell metadata information.

## Usage

```
annotateSpatialNetwork(
  gobject,
  spatial_network_name = "Delaunay_network",
  cluster_column,
  create_full_network = F
)
```

## **Arguments**

#### Value

annotated network in data.table format

# **Examples**

```
annotateSpatialNetwork(gobject)
```

# Description

annotate spatial locations with 2D spatial grid information

```
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

```
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
)
```

# **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)
```

binSpect binSpect

# Description

Previously: binGetSpatialGenes. BinSpect (Binary Spatial Extraction of genes) is a fast computational method that identifies genes with a spatially coherent expression pattern.

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#### Usage

```
binSpect(
  gobject,
  bin_method = c("kmeans", "rank"),
  expression_values = c("normalized", "scaled", "custom"),
  subset_genes = NULL,
  spatial_network_name = "Delaunay_network",
  nstart = 3,
  iter_max = 10,
  percentage_rank = 30,
  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
)
```

## **Arguments**

```
gobject
                  giotto object
                  method to binarize gene expression
bin_method
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
                  kmeans: iter.max parameter
iter_max
percentage_rank
                  percentage of top cells for binarization
do_fisher_test perform fisher test
calc_hub
                  calculate the number of hub cells
                  minimum number of cell-cell interactions for a hub cell
hub_min_int
                  calculate the average expression per gene of the high expressing cells
get_av_expr
                  calculate the number of high expressing cells per gene
get_high_expr
                  run calculations in parallel with mclapply
do_parallel
                  number of cores to use if do_parallel = TRUE
cores
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

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2. network: Alll cells are connected through a spatial network based on the physical coordinates

- 3. contingency table: A contingency table is calculated based on all edges of neighboring cells and the binarized expression (0-0, 0-1, 1-0 or 1-1)
- 4. For each gene an odds-ratio (OR) and fisher.test (optional) is calculated

Other statistics are provided (optional):

- Number of cells with high expression (binary = 1)
- Average expression of each gene within high expressing cells
- Number of hub cells, these are high expressing cells that have a user defined number of high expressing neighbors

By selecting a subset of likely spatial genes (e.g. soft thresholding highly variable genes) or using multiple cores can accelerate the speed.

#### Value

data.table with results (see details)

#### **Examples**

```
binSpect(gobject)
```

calculateHVG

calculateHVG

## **Description**

compute highly variable genes

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

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#### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
                  method to calculate highly variable genes
method
reverse_log_scale
                  reverse log-scale of expression values (default = FALSE)
logbase
                  if reverse_log_scale is TRUE, which log base was used?
expression_threshold
                  expression threshold to consider a gene detected
nr_expression_groups
                  number of expression groups for cov_groups
zscore_threshold
                  zscore to select hvg for cov_groups
HVGname
                  name for highly variable genes in cell metadata
difference_in_cov
                  minimum difference in coefficient of variance required
                  show plot
show_plot
                  return ggplot object
return_plot
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
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)

```
# 1. create giotto object
expr_path = system.file("extdata", "seqfish_field_expr.txt", package = 'Giotto')
loc_path = system.file("extdata", "seqfish_field_locs.txt", package = 'Giotto')
VC_small <- createGiottoObject(raw_exprs = expr_path, spatial_locs = loc_path)</pre>
```

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```
# 2. normalize giotto
VC_small <- normalizeGiotto(gobject = VC_small, scalefactor = 6000)
VC_small <- addStatistics(gobject = VC_small)
# 3. highly variable genes detection
VC_small <- calculateHVG(gobject = VC_small)</pre>
```

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

```
calculateMetaTable(gobject)
```

```
calculateMetaTableCells
```

calculateMetaTableCells

## **Description**

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

## Usage

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

## **Arguments**

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

## Value

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

## **Examples**

```
calculateMetaTableCells(gobject)
```

# Description

```
calculate_distance_and_weight
```

```
calculate_distance_and_weight(
  networkDT,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  d2_or_d3 = c(2, 3)
)
```

cellProximityBarplot 29

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

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

#### Value

```
ggplot barplot
```

```
cellProximityBarplot(CPscore)
```

```
cell Proximity Enrichment \\ cell Proximity Enrichment
```

# Description

Compute cell-cell interaction enrichment (observed vs expected)

## Usage

## **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.

```
cellProximityEnrichment(gobject)
```

```
cellProximityHeatmap cellProximityHeatmap
```

## **Description**

Create heatmap from cell-cell proximity scores

## Usage

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

## **Arguments**

```
giotto object
gobject
                  CPscore, output from cellProximityEnrichment()
CPscore
scale
                  scale cell-cell proximity interaction scores
order_cell_types
                  order cell types based on enrichment correlation
color_breaks
                  numerical vector of length 3 to represent min, mean and maximum
                  character color vector of length 3
color_names
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

#### **Details**

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

# Value

```
ggplot heatmap
```

#### **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
```

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```
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 = "Delaunay_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 Spat Plot 3D \\ cell Proximity Spat Plot 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 = "Delaunay_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**

cell Proximity SpatPlot 3D (gobject)

cellProximityVisPlot 39

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 = "Delaunay_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 = "Delaunay_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 = "Delaunay_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 = "Delaunay_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)
```

 ${\tt changeGiottoInstructions}$ 

change Giot to Instructions

# Description

Function to change one or more instructions from giotto object

# Usage

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

# Arguments

```
gobject giotto object

params parameter(s) to change

new_values new value(s) for parameter(s)

return_gobject (boolean) return giotto object
```

### Value

giotto object with one or more changed instructions

# **Examples**

changeGiottoInstructions()

changeImageBg 47

changeImageBg	changeImageBg
---------------	---------------

## **Description**

Function to change the background color of a magick image plot to another color

### Usage

```
changeImageBg(
  mg_object,
  bg_color,
  perc_range = 10,
  new_color = "#FFFFFF",
  new_name = NULL
)
```

# **Arguments**

mg\_object magick image or giotto image object
bg\_color estimated current background color
perc\_range range around estimated background color to include (percentage)

new\_color new background color

### Value

magick image or giotto image object with updated background color

### **Examples**

```
changeImageBg(mg_object)
```

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,
```

48 clusterCells

```
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,
     seed_number = 1234
   )
Arguments
   gobject
                   giotto object
   cluster_method community cluster method to use
                   name for new clustering result
   nn_network_to_use
                   type of NN network to use (kNN vs sNN)
   network_name
                   name of NN network to use
   \verb"pyth_leid_resolution"
                   resolution for leiden
   pyth_leid_weight_col
                   column to use for weights
   pyth_leid_part_type
```

partition type to use

clusterCells 49

pyth\_leid\_init\_memb

initial membership

pyth\_leid\_iterations

number of iterations

pyth\_louv\_resolution

resolution for louvain

pyth\_louv\_weight\_col

python louvain param: weight column

python\_louv\_random

python louvain param: random

python\_path specify specific path to python if required louvain\_gamma louvain param: gamma or resolution

louvain\_omega louvain param: omega

walk\_steps randomwalk: number of steps
walk\_clusters randomwalk: number of clusters
walk\_weights randomwalk: weight column
sNNclust\_k SNNclust: k neighbors to use

sNNclust\_eps SNNclust: epsilon

sNNclust\_minPts

SNNclust: min points

borderPoints SNNclust: border points

expression\_values

expression values to use

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

dimension reduction to use

dim\_reduction\_name

name of reduction 'pca',

dimensions\_to\_use

dimensions to use

distance\_method

distance method

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

km\_nstart kmeans random starting points

km\_algorithm kmeans algorithm

hc\_agglomeration\_method

hierarchical clustering method

hc\_k hierachical number of clusters

hc\_h hierarchical tree cutoff

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

set\_seed set seed

seed\_number number for seed

### **Details**

Wrapper for the different clustering methods.

### Value

giotto object with new clusters appended to cell metadata

### See Also

 $\label{local-community} do Louvain Cluster\_community, do Louvain Cluster\_multinet, do Louvain Cluster, do Random Walk Cluster, do SNN Cluster, do Kmeans, do H clust$ 

## **Examples**

```
clusterCells(gobject)
```

clusterSpatialCorGenes

cluster Spatial Cor Genes

# Description

Cluster based on spatially correlated genes

## Usage

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

# **Arguments**

spatCorObject spatial correlation object

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

return\_obj return spatial correlation object (spatCorObject)

### Value

spatCorObject or cluster results

## **Examples**

clusterSpatialCorGenes(gobject)

colMeans\_giotto 51

```
colMeans_giotto colMeans_giotto
```

## **Description**

```
colMeans_giotto
```

# Usage

```
colMeans_giotto(mymatrix)
```

colSums\_giotto

colSums\_giotto

## Description

```
colSums_giotto
```

# Usage

```
colSums_giotto(mymatrix)
```

combCCcom

combCCcom

## **Description**

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

# Usage

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

```
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

## **Examples**

```
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
)
```

```
cpgObject
                  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
                  minimum number of interacting cell type
min_int_cells
min_fdr
                  minimum adjusted p-value
                  minimum absolute spatial expression difference
min_spat_diff
                  minimum absolute log2 fold-change
min_log2_fc
do_parallel
                  run calculations in parallel with mclapply
cores
                  number of cores to use if do_parallel = TRUE
verbose
                  verbose
```

## Value

cpgObject that contains the filtered differential gene scores

# **Examples**

```
combineCellProximityGenes(gobject)
```

```
combine {\tt CellProximityGenes\_per\_interaction} \\ combine {\tt CellProximityGenes\_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

```
{\tt combineCellProximityGenes\_per\_interaction()}
```

combineCPG

combineCPG

# Description

Combine CPG scores in a pairwise manner.

54 combineCPG

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

## **Arguments**

```
cpgObject
                  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
                  minimum absolute spatial expression difference
min_spat_diff
                  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 55

combineMetadata

combineMetadata

## **Description**

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

## Usage

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

# **Arguments**

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

### Value

Extended cell metadata in data.table format.

## **Examples**

```
combineMetadata(gobject)
```

```
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**

converts a magick image object to a data.table

# Usage

```
convert_mgImage_to_array_DT(mg_object)
```

# Arguments

mg\_object magick image or Giotto image object

### Value

data.table with image pixel information

```
convert\_to\_full\_spatial\_network \\ convert\_to\_full\_spatial\_network
```

# Description

convert to a full spatial network

## Usage

```
convert_to_full_spatial_network(reduced_spatial_network_DT)
```

## **Description**

convert to a reduced spatial network

# Usage

```
convert_to_reduced_spatial_network(full_spatial_network_DT)
```

cor\_giotto 57

cor\_giotto

cor\_giotto

## Description

```
cor_giotto
```

## Usage

```
cor_giotto(x, ...)
```

cor\_sparse

cor\_sparse adapted from wydr package

# Description

cor\_sparse adapted from wydr package

## Usage

```
cor_sparse(x)
```

createCrossSection

*createCrossSection* 

# Description

Create a virtual 2D cross section.

## Usage

```
createCrossSection(
  gobject,
  name = "cross_section",
  spatial_network_name = "Delaunay_network",
  thickness_unit = c("cell", "natural"),
  slice_thickness = 2,
  cell_distance_estimate_method = "mean",
  extend_ratio = 0.2,
  method = c("equation", "3 points", "point and norm vector",
    "point and two plane vectors"),
  equation = NULL,
  point1 = NULL,
  point2 = NULL,
  point3 = NULL,
  normVector = NULL,
  planeVector1 = NULL,
  planeVector2 = NULL,
  mesh\_grid\_n = 20,
  return\_gobject = TRUE
```

58 createCrossSection

### **Arguments**

gobject giotto object

name name of cress section object. (default = cross\_sectino)

spatial\_network\_name

name of spatial network object. (default = Delaunay\_network)

thickness\_unit unit of the virtual section thickness. If "cell", average size of the observed

cells is used as length unit. If "natural", the unit of cell location coordinates

is used.(default = cell)

 ${\tt cell\_distance\_estimate\_method}$ 

method to estimate average distance between neighboring cells. (default = mean)

extend\_ratio deciding the span of the cross section meshgrid, as a ratio of extension compared

to the borders of the vitural tissue section. (default = 0.2)

method method to define the cross section plane. If equation, the plane is defined by

a four element numerical vector (equation) in the form of c(A,B,C,D), corresponding to a plane with equation Ax+By+Cz=D. If 3 points, the plane is define by the coordinates of 3 points, as given by point1, point2, and point3. If point and norm vector, the plane is defined by the coordinates of one point (point1) in the plane and the coordinates of one norm vector (normVector) to the plane. If point and two plane vector, the plane is defined by the coordinates of one point (point1) in the plane and the coordinates of two vectors (planeVector1,

planeVector2) in the plane. (default = equation)

equation equation required by method "equation".equations needs to be a numerical vec-

tor of length 4, in the form of c(A,B,C,D), which defines plane Ax+By+Cz=D.

point1 coordinates of the first point required by method "3 points", "point and norm

vector", and "point and two plane vectors".

point2 coordinates of the second point required by method "3 points"

point3 coordinates of the third point required by method "3 points"

normVector coordinates of the norm vector required by method "point and norm vector"

planeVector1 coordinates of the first plane vector required by method "point and two plane

vectors"

planeVector2 coordinates of the second plane vector required by method "point and two plane

vectors"

mesh\_grid\_n numer of meshgrid lines to generate along both directions for the cross section

plane.

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

#### **Details**

Creates a virtual 2D cross section object for a given spatial network object. The users need to provide the definition of the cross section plane (see method).

## Value

giotto object with updated spatial network slot

createGiottoImage 59

createGiottoImage

# Description

Creates a giotto image that can be added to a Giotto object and/or used to add an image to the spatial plotting functions

# Usage

```
createGiottoImage(
  gobject = NULL,
  spatial_locs = NULL,
  mg_object,
  name = "image",
  xmax_adj = 0,
  xmin_adj = 0,
  ymax_adj = 0,
  ymin_adj = 0
)
```

# **Arguments**

```
giotto object
gobject
spatial_locs
                  spatial locations (alternative if giobject = NULL)
mg_object
                  magick image object
name
                  name for the image
xmax_adj
                  adjustment of the maximum x-value to align the image
xmin_adj
                  adjustment of the minimum x-value to align the image
                  adjustment of the maximum y-value to align the image
ymax_adj
ymin_adj
                  adjustment of the minimum y-value to align the image
```

### Value

```
a giotto image object
```

# **Examples**

```
createGiottoImage(mg_object)
```

60 createGiottoInstructions

```
{\tt 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
```

# Arguments

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

### Value

named vector with giotto instructions

# **Examples**

```
createGiottoInstructions()
```

createGiottoObject 61

### **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,
  images = NULL,
  offset_file = NULL,
  instructions = NULL,
  cores = NA
```

```
matrix with raw expression counts [required]
raw_exprs
spatial_locs
                  data.table or data.frame with coordinates for cell centroids
                  normalized expression values
norm_expr
norm_scaled_expr
                  scaled expression values
custom_expr
                  custom expression values
cell_metadata
                  cell annotation metadata
gene_metadata
                  gene annotation metadata
spatial_network
                  list of spatial network(s)
spatial_network_name
                  list of spatial network name(s)
spatial_grid
                  list of spatial grid(s)
spatial_grid_name
                  list of spatial grid name(s)
```

62 createGiottoObject

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)

nn\_network list of nearest neighbor network(s)

images list of images

offset\_file file used to stitch fields together (optional)

instructions list of instructions or output result from createGiottoInstructions cores how many cores or threads to use to read data if paths are provided

#### **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. This matrix can be provided as a base matrix, sparse Matrix, data.frame, data.table or as a path to any of those. To include spatial information about cells (or regions) you need to provide a matrix, data.table or data.frame (or path to them) 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
- images

### Value

giotto object

## **Examples**

createGiottoObject(raw\_exprs, spatial\_locs)

```
create {\it Giotto Visium Object} \\ {\it create Giotto Visium Object}
```

### **Description**

creates Giotto object directly from a 10X visium folder

### Usage

```
createGiottoVisiumObject(
  visium_dir = NULL,
  expr_data = c("raw", "filter"),
  gene_column_index = 1,
  png_name = NULL,
  xmax_adj = 0,
  xmin_adj = 0,
  ymax_adj = 0,
  ymin_adj = 0,
  instructions = NULL,
  cores = NA
)
```

### **Arguments**

```
visium_dir
                  path to the 10X visium directory [required]
expr_data
                  raw or filtered data (see details)
gene_column_index
                  which column index to select (see details)
                  select name of png to use (see details)
png_name
xmax_adj
                  adjustment of the maximum x-value to align the image
                  adjustment of the minimum x-value to align the image
xmin_adj
ymax_adj
                  adjustment of the maximum y-value to align the image
ymin_adj
                  adjustment of the minimum y-value to align the image
instructions
                  list of instructions or output result from createGiottoInstructions
                  how many cores or threads to use to read data if paths are provided
cores
```

## **Details**

- expr\_data: raw will take expression data from raw\_feature\_bc\_matrix and filter from filtered\_feature\_bc\_matrix
- gene\_column\_index: which gene identifiers (names) to use if there are multiple columns (e.g. ensemble and gene symbol)
- png\_name: by default the first png will be selected, provide the png name to override this (e.g. myimage.png)

# Value

giotto object

64 createHeatmap\_DT

#### **Examples**

```
createGiottoVisiumObject(visium_dir)
```

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("correlation", "custom"),
  gene_custom_order = NULL,
  gene_cor_method = "pearson",
  gene_hclust_method = "complete"
)
```

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

createMetagenes 65

### **Details**

Creates input data.tables for plotHeatmap function.

### Value

list

# **Examples**

```
createHeatmap_DT(gobject)
```

createMetagenes

createMetagenes

# **Description**

This function creates an average metagene for gene clusters.

### Usage

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

## **Arguments**

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

## Details

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

#### Value

giotto object

## **Examples**

```
createMetagenes(gobject)
```

66 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
. . .
```

#### **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)

 ${\tt createSpatialDelaunayNetwork}$ 

create Spatial Delaunay Network

## **Description**

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

#### Usage

```
createSpatialDelaunayNetwork(
  gobject,
  method = c("deldir", "delaunayn_geometry", "RTriangle"),
  dimensions = "all",
  name = "delaunay_network",
  maximum_distance = "auto",
  minimum_k = 0,
  options = "Pp",
  Y = TRUE,
  j = TRUE,
  S = 0,
  verbose = T,
  return_gobject = TRUE,
  ...
)
```

### **Arguments**

gobject giotto object dimensions which spatial dimensions to use (default = all) name name for spatial network (default = 'delaunay\_network') maximum\_distance distance cuttof for Delaunay neighbors to consider. If "auto", "upper wisker" value of the distance vector between neighbors is used; see the boxplotgraphics documentation for more details.(default = "auto") minimum number of neighbours if maximum\_distance != NULL minimum\_k options (geometry) String containing extra control options for the underlying Qhull command; see the Qhull documentation (../doc/qhull/html/qdelaun.html) for the available options. (default = 'Pp', do not report precision problems) Υ (RTriangle) If TRUE prohibits the insertion of Steiner points on the mesh boundary. (RTriangle) If TRUE jettisons vertices that are not part of the final triangulation j from the output. (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 delaunayn (default), deldir, or triangulate.

## Value

giotto object with updated spatial network slot

# **Examples**

```
createSpatialDelaunayNetwork(gobject)
```

createSpatialEnrich 69

createSpatialEnrich createSpatialEnrich

## **Description**

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

## Usage

```
createSpatialEnrich(
  gobject,
  enrich_method = c("PAGE", "rank", "hypergeometric"),
  sign_matrix,
  expression_values = c("normalized", "scaled", "custom"),
  reverse_log_scale = TRUE,
  logbase = 2,
  p_value = TRUE,
  n_genes = 100,
  n_times = 1000,
  top_percentage = 5,
  output_enrichment = c("original", "zscore"),
  name = "PAGE",
  return_gobject = TRUE
)
```

```
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
logbase
                  log base to use if reverse_log_scale = TRUE
p_value
                  calculate p-value (default = FALSE)
n_times
                  (page/rank) number of permutation iterations to calculate p-value
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
```

70 createSpatialGrid

#### **Details**

For details see the individual functions:

```
PAGE: PAGEEnrichPAGE: rankEnrichPAGE: hyperGeometricEnrich
```

### Value

Giotto object or enrichment results if return\_gobject = FALSE

### **Examples**

```
createSpatialEnrich(gobject)
```

createSpatialGrid

createSpatialGrid

## **Description**

Create a spatial grid.

# Usage

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

# **Arguments**

```
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.

createSpatialGrid\_2D 71

### Value

giotto object with updated spatial grid slot

## **Examples**

```
createSpatialGrid(gobject)
```

```
createSpatialGrid\_2D \quad \textit{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

## **Examples**

```
createSpatialGrid_2D(gobject)
```

72 createSpatialGrid\_3D

```
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

## **Examples**

```
createSpatialGrid_3D(gobject)
```

```
createSpatialKNNnetwork
```

*createSpatialKNNnetwork* 

#### **Description**

Create a spatial knn network.

#### Usage

```
createSpatialKNNnetwork(
  gobject,
  method = "dbscan",
  dimensions = "all",
  name = "knn_network",
  k = 4,
  maximum_distance = NULL,
  minimum_k = 0,
  verbose = F,
  return_gobject = TRUE,
  ...
)
```

### **Arguments**

gobject giotto object

method method to create kNN network

dimensions which spatial dimensions to use (default = all)

name name for spatial network (default = 'spatial\_network')k number of nearest neighbors based on physical distance

maximum\_distance

distance cuttof for nearest neighbors to consider for kNN network

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

verbose verbose

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

## Value

giotto object with updated spatial network slot

**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

#### **Examples**

```
createSpatialKNNnetwork(gobject)
```

74 createSpatialNetwork

```
create Spatial Network \\ create Spatial Network
```

# Description

Create a spatial network based on cell centroid physical distances.

# Usage

```
createSpatialNetwork(
  gobject,
  name = NULL,
  dimensions = "all",
  method = c("Delaunay", "kNN"),
delaunay_method = c("deldir", "delaunayn_geometry", "RTriangle"),
maximum_distance_delaunay = "auto",
  options = "Pp",
  Y = TRUE,
  j = TRUE,
  S = 0,
  minimum_k = 0,
  knn_method = "dbscan",
  k = 4,
  maximum_distance_knn = NULL,
  verbose = F,
  return_gobject = TRUE,
)
```

# Arguments

gobject	giotto object
name	name for spatial network (default = 'spatial_network')
dimensions	which spatial dimensions to use (default = all)
method	which method to use to create a spatial network. (default = Delaunay)
delaunay_method	
	Delaunay method to use
maximum_distance_delaunay	
	distance cuttof for nearest neighbors to consider for Delaunay network
options	(geometry) String containing extra control options for the underlying Qhull command; see the Qhull documentation (/doc/qhull/html/qdelaun.html) for the available options. (default = 'Pp', do not report precision problems)
Υ	(RTriangle) If TRUE prohibits the insertion of Steiner points on the mesh boundary.
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.
minimum_k	minimum nearest neigbhours if maximum_distance != NULL

knn\_method method to create kNN network

k number of nearest neighbors based on physical distance

maximum\_distance\_knn

distance cuttof for nearest neighbors to consider for kNN 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. For Delaunay method, neighbors will be decided by delaunay triangulation and a maximum distance criteria. For kNN method, number of neighbors can be determined by k, or 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

#### Value

giotto object with updated spatial network slot

# **Examples**

createSpatialNetwork(gobject)

## **Description**

create 2d mesh grid line object

```
create_2d_mesh_grid_line_obj(x_min, x_max, y_min, y_max, mesh_grid_n)
```

76 create\_average\_DT

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

## **Description**

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

## Usage

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

### **Arguments**

# Value

data.table with average gene epression values for each factor

# **Description**

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

## Usage

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

# Arguments

```
gobject giotto object

meta_data_name name of metadata column to use
expression_values

which expression values to use
```

#### Value

data.table with average gene epression values for each factor

# Description

creates randomized cell ids within a selection of cell types

# Usage

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

## **Arguments**

```
gobject giotto object to use

cluster_column cluster column with cell type information

needed_cell_types

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

# **Details**

Details will follow.

## Value

list of randomly sampled cell ids with same cell type composition

## **Examples**

```
create_cell_type_random_cell_IDs(gobject)
```

```
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\_crossSection\_object \\ create\_crossSection\_object
```

# Description

create a crossSection object

```
create_crossSection_object(
  name = NULL,
  method = NULL,
  thickness_unit = NULL,
  slice_thickness = NULL,
  plane_equation = NULL,
  mesh_grid_n = NULL,
  mesh_obj = NULL,
  cell_subset = NULL,
  cell_subset_spatial_locations = NULL,
  cell_subset_projection_locations = NULL,
  cell_subset_projection_PCA = NULL,
  cell_subset_projection_coords = NULL)
```

```
create_delaunayNetwork2D
```

create\_delaunayNetwork2D

# Description

Create a spatial Delaunay network.

## Usage

```
create_delaunayNetwork2D(
  gobject,
  method = c("delaunayn_geometry", "RTriangle", "deldir"),
  sdimx = "sdimx",
  sdimy = "sdimy",
  name = "delaunay_network",
  maximum_distance = "auto",
  minimum_k = 0,
  options = "Pp",
  Y = TRUE,
  j = TRUE,
  S = 0,
  verbose = T,
  return_gobject = TRUE,
  ...
)
```

## **Examples**

 ${\tt create\_delaunayNetwork2D(gobject)}$ 

```
create_delaunayNetwork3D
```

create\_delaunayNetwork3D

# Description

Create a spatial Delaunay network.

```
create_delaunayNetwork3D(
  gobject,
  method = "delaunayn_geometry",
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  name = "delaunay_network_3D",
  maximum_distance = "auto",
```

```
minimum_k = 0,
options = "Pp",
return_gobject = TRUE,
...
)
```

## **Examples**

```
create_delaunayNetwork3D(gobject)
```

## **Description**

Create a spatial Delaunay network.

# Usage

```
create_delaunayNetwork_deldir(
  spatial_locations,
  sdimx = "sdimx",
  sdimy = "sdimy",
   ...
)
```

## **Examples**

 ${\tt create\_delaunayNetwork\_deldir(gobject)}$ 

```
create\_delaunayNetwork\_geometry \\ create\_delaunayNetwork\_geometry
```

# Description

Create a spatial Delaunay network.

# Usage

```
create_delaunayNetwork_geometry(
  spatial_locations,
  sdimx = "sdimx",
  sdimy = "sdimy",
  options = "Pp",
   ...
)
```

# **Examples**

```
{\tt create\_delaunayNetwork\_geometry(gobject)}
```

```
\label{lem:create_delaunayNetwork_geometry_3D} create\_delaunayNetwork\_geometry\_3D
```

# Description

Create a spatial Delaunay network.

# Usage

```
create_delaunayNetwork_geometry_3D(
   spatial_locations,
   sdimx = "sdimx",
   sdimy = "sdimy",
   sdimz = "sdimz",
   options = options,
   ...
)
```

# **Examples**

```
create_delaunayNetwork_geometry_3D(gobject)
```

```
create\_delaunayNetwork\_RTriangle \\ create\_delaunayNetwork\_RTriangle
```

## **Description**

Create a spatial Delaunay network.

# Usage

```
create_delaunayNetwork_RTriangle(
   spatial_locations,
   sdimx = "sdimx",
   sdimy = "sdimy",
   Y = TRUE,
   j = TRUE,
   S = 0,
   ...
)
```

## **Examples**

```
create_delaunayNetwork_RTriangle(gobject)
```

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

 $\begin{array}{ll} \text{name} & \text{arbitrary name for object} \\ \text{reduction\_method} \end{array}$ 

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

## **Description**

subsets matrix based on vector of genes or hvg column

## Usage

```
create_genes_to_use_matrix(gobject, sel_matrix, genes_to_use, verbose = TRUE)
```

# Arguments

gobject giotto object

sel\_matrix selected expression matrix

genes\_to\_use genes to use, character or vector of genes

verbose verbosity

# Value

subsetted matrix based on selected genes

create\_jackstrawplot 83

```
create_jackstrawplot create_jackstrawplot
```

# Description

create jackstrawplot with ggplot

## Usage

```
create_jackstrawplot(
  jackstraw_data,
  ncp = 20,
  ylim = c(0, 1),
  threshold = 0.01
)
```

## Arguments

```
jackstraw_data result from jackstraw function
ncp number of principal components to calculate
ylim y-axis limits on jackstraw plot
p-value threshold to call a PC significant
```

# Value

ggplot

```
create\_KNNnetwork\_dbscan \\ create\_KNNnetwork\_dbscan
```

## Description

Create a spatial knn network.

## Usage

```
create_KNNnetwork_dbscan(
   spatial_locations,
   sdimx = "sdimx",
   sdimy = "sdimy",
   sdimz = "sdimz",
   k = 4,
   ...
)
```

## **Examples**

```
create_KNNnetwork_dbscan(gobject)
```

84 create\_screeplot

# Description

create mesh grid lines for cross section

# Usage

```
create_mesh_grid_lines(
  cell_subset_projection_locations,
  extend_ratio,
  mesh_grid_n
)
```

create\_screeplot

create\_screeplot

# Description

create screeplot with ggplot

# Usage

```
create_screeplot(pca_obj, ncp = 20, ylim = c(0, 20))
```

# Arguments

pca\_obj pca dimension reduction object

ncp number of principal components to calculate

ylim y-axis limits on scree plot

# Value

ggplot

## **Description**

creates a spatial network object to store the created spatial network and additional information

## Usage

```
create_spatialNetworkObject(
  name = NULL,
  method = NULL,
  parameters = NULL,
  outputObj = NULL,
  networkDT = NULL,
  cellShapeObj = NULL,
  networkDT_before_filter = NULL,
  crossSectionObjects = NULL,
  misc = NULL
)
```

crossSectionGenePlot crossSectionGenePlot

# Description

Visualize cells and gene expression in a virtual cross section according to spatial coordinates

```
crossSectionGenePlot(
  gobject = NULL,
  crossSection_obj = NULL,
  name = NULL,
  spatial_network_name = "Delaunay_network",
  expression_values = c("normalized", "scaled", "custom"),
  genes_high_color = "red",
  genes_mid_color = "white",
  genes_low_color = "darkblue",
  show_network = F,
  network_color = NULL,
  edge_alpha = NULL,
  show_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  midpoint = 0,
  scale_alpha_with_expression = FALSE,
```

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```
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 = "crossSectionGenePlot"
    )
Arguments
    gobject
                     giotto object
                     name of virtual cross section to use
    name
    spatial_network_name
                     name of spatial network to use
    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
    network_color
                     color of spatial network
    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
    point_size
                     size of point (cell)
    point_border_col
```

color of border around points

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```
point_border_stroke
                 stroke size of border around points
show_legend
                 show legend
                 size of legend text
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
                 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
```

## **Details**

Description of parameters.

#### Value

ggplot

#### See Also

spatGenePlot3D and spatGenePlot2D

# **Examples**

crossSectionGenePlot(gobject)

crossSectionGenePlot3D

crossSectionGenePlot3D

parameters for cowplot::save\_plot()

# Description

Visualize cells and gene expression in a virtual cross section according to spatial coordinates

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#### Usage

```
crossSectionGenePlot3D(
  gobject,
  crossSection_obj = NULL,
  name = NULL,
  spatial_network_name = "Delaunay_network",
  expression_values = c("normalized", "scaled", "custom"),
  genes,
  show_network = F,
  network_color = NULL,
  edge_alpha = NULL,
  show\_grid = F,
  cluster_column = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
  other_cell_color = alpha("lightgrey", 0),
  other_point_size = 1,
  genes_high_color = "red",
  genes_mid_color = "white",
  genes_low_color = "darkblue",
  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 = "crossSectionGenePlot3D"
)
```

# **Arguments**

```
gobject
                  giotto object
                  name of virtual cross section to use
name
spatial_network_name
                  name of spatial network to use
expression_values
                  gene expression values to use
                  genes to show
genes
show_network
                  show underlying spatial network
network_color
                  color of spatial network
show_grid
                  show spatial grid
genes_high_color
                  color represents high gene expression
```

```
genes_mid_color
                  color represents middle gene expression
genes_low_color
                  color represents low gene expression
spatial_grid_name
                  name of spatial grid to use
                  size of point (cell)
point_size
show_legend
                  show legend
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
grid_color
                  color of spatial grid
                  expression midpoint
midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  parameters for cowplot::save_plot()
. . .
```

#### **Details**

Description of parameters.

### Value

ggplot

## **Examples**

crossSectionGenePlot3D(gobject)

# Description

Visualize cells in a virtual cross section according to spatial coordinates

```
crossSectionPlot(
  gobject,
  crossSection_obj = NULL,
  name = NULL,
  spatial_network_name = "Delaunay_network",
  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,
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,
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 = "crossSectionPlot"
```

#### **Arguments**

gobject giotto object name of virtual cross section to use name spatial\_network\_name name of spatial network to use 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) 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\_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 underlying spatial network show\_network 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
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
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
                  create multiple plots based on cell annotation column
groub_by
```

### **Details**

Description of parameters.

#### Value

ggplot

## See Also

crossSectionPlot

crossSectionPlot3D crossSectionPlot3D

## **Description**

Visualize cells in a virtual cross section according to spatial coordinates

### Usage

```
crossSectionPlot3D(
  gobject,
  crossSection_obj = NULL,
  name = NULL,
  spatial_network_name = "Delaunay_network",
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  point_size = 3,
  cell_color = NULL,
  cell_color_code = NULL,
  select_cell_groups = NULL,
  show_other_cells = T,
  other_cell_color = alpha("lightgrey", 0),
  other_point_size = 0.5,
  show_network = F,
  network_color = NULL,
  network\_alpha = 1,
  other_cell_alpha = 0.5,
  show\_grid = F,
  grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  title = "",
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
  x_ticks = NULL,
  y_ticks = NULL,
  z_ticks = NULL,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "crossSection3D"
)
```

### **Arguments**

```
gobject giotto object

name name of virtual cross section to use
spatial_network_name

name of spatial network to use
```

```
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
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
network_color
                  color of spatial network
show_grid
                  show spatial grid
grid_color
                  color of spatial grid
spatial_grid_name
                  name of spatial grid to use
title
                  title of plot
show_legend
                  show legend
axis_scale
                  the way to scale the axis
custom_ratio
                  customize the scale of the plot
                  set the number of ticks on the x-axis
x_ticks
y_ticks
                  set the number of ticks on the y-axis
z_ticks
                  set the number of ticks on the z-axis
show_plot
                  show plot
return_plot
                  return ggplot object
                  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**

```
{\tt crossSectionPlot3D(gobject)}
```

decide\_cluster\_order 95

```
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

## **Examples**

```
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 = "Delaunay_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()

detectSpatialPatterns 97

#### Value

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

### See Also

```
showSpatialCorGenes
```

## **Examples**

```
detectSpatialCorGenes(gobject)
```

```
detectSpatialPatterns detectSpatialPatterns
```

## **Description**

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

# Usage

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

## Arguments

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

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#### **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_alpha = 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
    gobject
                     giotto object
   dim_reduction_to_use
                     dimension reduction to use
   dim_reduction_name
                     dimension reduction name
   dim1_to_use
                     dimension to use on x-axis
   dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    show_NN_network
                     show underlying NN network
   nn\_network\_to\_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
   network_name
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
```

select 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
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
                  column to use for alpha of the edges
edge_alpha
                  point with border or not (border or no_border)
point_shape
                  size of point (cell)
point_size
                  transparancy of dim. reduction points
point_alpha
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
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
                  title for plot, defaults to cell_color parameter
title
```

# Details

Description of parameters. For 3D plots see dimCellPlot2D

dimCellPlot2D

#### Value

ggplot

#### **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_alpha = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
```

dimCellPlot2D

```
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 = "dimCellPlot2D"
    )
Arguments
    gobject
                     giotto object
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                     numeric cell annotation columns
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
    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
```

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```
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_alpha
                  transparancy of dim. reduction points
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
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
                  title for plot, defaults to cell_color parameter
title
```

# **Details**

Description of parameters. For 3D plots see dimPlot3D

### Value

ggplot

# Examples

dimCellPlot2D(gobject)

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 ${\tt dimGenePlot}$ 

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,
  point_alpha = 1,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  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 = "dimGenePlot"
)
```

## Arguments

gobject giotto object

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```
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
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)
                 transparancy of points
point_alpha
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                 midpoint for color gradient
gradient_limits
                 vector with lower and upper limits
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
                 cowplot param: relative height
cow_rel_h
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]
                 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.

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#### Value

ggplot

#### See Also

```
dimGenePlot3D
```

#### **Examples**

```
dimGenePlot(gobject)
```

dimGenePlot2D

dimGenePlot2D

## **Description**

Visualize cells and gene expression according to dimension reduction coordinates

```
dimGenePlot2D(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  scale_alpha_with_expression = FALSE,
  point_shape = c("border", "no_border"),
  point_size = 1,
  point_alpha = 1,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  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",
```

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```
show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot2D"
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
                     genes to show
    genes
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
    dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
                     column to use for alpha of the edges
    edge_alpha
    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_alpha
                     transparancy of points
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    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
                     size of axis title
    axis_title
    cow_n_col
                     cowplot param: how many columns
```

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```
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
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  parameters for cowplot::save_plot()
```

#### **Details**

Description of parameters.

#### Value

ggplot

#### See Also

dimGenePlot3D

#### **Examples**

```
dimGenePlot2D(gobject)
```

dimGenePlot3D

dimGenePlot3D

## **Description**

Visualize cells and gene expression according to dimension reduction coordinates

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

dimGenePlot3D

```
select_cell_groups = NULL,
      select_cells = NULL,
      show_other_cells = T,
      other_cell_color = "lightgrey",
      other_point_size = 1,
      edge_alpha = NULL,
      point_size = 2,
      genes_high_color = NULL,
      genes_mid_color = "white",
      genes_low_color = "blue",
      show_legend = T,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot3D"
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
                     genes to show
    genes
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
    dim1_to_use
                     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
    edge_alpha
                     column to use for alpha of the edges
                     size of point (cell)
    point_size
    show_legend
                     show legend
    show_plot
                     show plots
    return_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()

. . .

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#### **Details**

Description of parameters.

#### Value

ggplot

# **Examples**

```
dimGenePlot3D(gobject)
```

dimPlot

dimPlot

# **Description**

Visualize cells according to dimension reduction coordinates

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

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```
point_shape = c("border", "no_border"),
      point_size = 1,
      point_alpha = 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,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimPlot"
    )
Arguments
    gobject
                    giotto object
    group_by_subset
                    subset the group_by factor column
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
    dim1_to_use
                    dimension to use on x-axis
                    dimension to use on y-axis
    dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                    show underlying NN network
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
   network_name
                    name of NN network to use, if show_NN_network = TRUE
    cell_color
                    color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                    named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient\_midpoint
```

midpoint for color gradient

dimPlot

```
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
                  column to use for alpha of the edges
edge_alpha
                  point with border or not (border or no_border)
point_shape
point_size
                  size of point (cell)
point_alpha
                  transparancy of point
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
axis_title
                  size of axis title
                  title for plot, defaults to cell_color parameter
title
                  cowplot param: how many columns
cow_n_col
                  cowplot param: relative height
cow_rel_h
cow_rel_w
                  cowplot param: relative width
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
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
                  create multiple plots based on cell annotation column
groub_by
```

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#### **Details**

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

#### Value

ggplot

# **Examples**

dimPlot(gobject)

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,
```

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```
point_shape = c("border", "no_border"),
      point_size = 1,
      point_alpha = 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,
      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
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
    dim1_to_use
                    dimension to use on x-axis
                    dimension to use on y-axis
    dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                    show underlying NN network
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
   network_name
                    name of NN network to use, if show_NN_network = TRUE
    cell_color
                    color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                    named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient\_midpoint
```

midpoint for color gradient

gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select 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 column to use for alpha of the edges edge\_alpha point with border or not (border or no\_border) point\_shape point\_size size of point (cell) point\_alpha transparancy of point 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 cowplot param: how many columns cow\_n\_col cowplot param: relative height cow\_rel\_h cow\_rel\_w cowplot param: relative width cowplot param: how to align cow\_align show\_plot show plot return\_plot return ggplot object save\_plot directly save the plot [boolean] 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 create multiple plots based on cell annotation column groub\_by

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#### **Details**

Description of parameters. For 3D plots see dimPlot3D

#### Value

ggplot

### **Examples**

```
dimPlot2D(gobject)
```

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,
```

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```
point_alpha = 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,
      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
                     dimension to use on x-axis
    dim1_to_use
                     dimension to use on y-axis
    dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     name of NN network to use, if show_NN_network = TRUE
    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

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

# **Details**

Description of parameters. For 3D plots see dimPlot3D

### Value

ggplot

# **Examples**

```
{\tt dimPlot2D\_single(gobject)}
```

dimPlot3D

dimPlot3D

dimPlot3D

### **Description**

Visualize cells according to dimension reduction coordinates

# 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"
)
```

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```
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_NN_network
                  show underlying NN network
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  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
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
                  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
{\sf show\_legend}
                  show legend
```

# **Details**

Description of parameters.

#### Value

plotly

# **Examples**

dimPlot3D(gobject)

doHclust 121

doHclust doHclust

# **Description**

cluster cells using hierarchical clustering algorithm

### Usage

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

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

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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

hclust

### **Examples**

```
doHclust(gobject)
```

doHMRF

doHMRF

### **Description**

Run HMRF

# Usage

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

doKmeans 123

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

### **Details**

Description of HMRF parameters ...

### Value

Creates a directory with results that can be viewed with viewHMRFresults

### **Examples**

```
doHMRF(gobject)
```

doKmeans

doKmeans

# **Description**

cluster cells using kmeans algorithm

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

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

### **Arguments**

gobject giotto object

expression\_values

expression values to use

genes\_to\_use subset of genes to use

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimensions reduction name

dimensions\_to\_use

dimensions to use

distance\_method

distance method

centers number of final clusters iter\_max kmeans maximum iterations

nstart kmeans nstart algorithm kmeans algorithm

name name for kmeans clustering

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

set\_seed set seed

seed\_number number for seed

### **Details**

Description on how to use Kmeans clustering method.

# Value

giotto object with new clusters appended to cell metadata

# See Also

kmeans

# **Examples**

doKmeans(gobject)

doLeidenCluster 125

doLeidenCluster doLeidenCluster

### **Description**

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

# Usage

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

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

126 doLeidenSubCluster

#### **Details**

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

Partition types available and information:

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

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

#### Value

giotto object with new clusters appended to cell metadata

### **Examples**

```
doLeidenCluster(gobject)
```

doLeidenSubCluster

doLeidenSubCluster

### **Description**

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

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

doLeidenSubCluster 127

```
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

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

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

resolution resolution of Leiden clustering

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

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

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

verbose verbose

# Details

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

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

# Value

giotto object with new subclusters appended to cell metadata

128 doLouvainCluster

### See Also

```
doLeidenCluster
```

### **Examples**

```
doLeidenSubCluster(gobject)
```

doLouvainCluster

doLouvainCluster

### **Description**

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

### Usage

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

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

### **Details**

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

#### Value

giotto object with new clusters appended to cell metadata

### See Also

doLouvainCluster\_community and doLouvainCluster\_multinet

# **Examples**

```
doLouvainCluster(gobject)
```

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

# **Description**

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

### Usage

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

weight\_col weight column to use for edges

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

to get different partitions at each call

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

set\_seed set seed

seed\_number number for seed

### **Details**

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

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

#### Value

giotto object with new clusters appended to cell metadata

### **Examples**

```
doLouvainCluster_community(gobject)
```

```
doLouvainCluster_multinet
```

doLouvainCluster\_multinet

# **Description**

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

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

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#### **Arguments**

seed\_number

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

### Value

**Details** 

giotto object with new clusters appended to cell metadata

number for seed

# **Examples**

```
doLouvainCluster_multinet(gobject)
```

```
doLouvainSubCluster doLouvainSubCluster
```

# Description

subcluster cells using a NN-network and the Louvain algorithm

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

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

### **Arguments**

gobject giotto object

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

cluster\_column cluster column to subcluster

selected\_clusters

only do subclustering on these clusters

hvg\_param parameters for calculateHVG

 $hvg\_min\_perc\_cells$ 

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

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

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

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork resolution resolution for community algorithm

gamma gamma omega omega

python\_path specify specific path to python if required

 $nn\_network\_to\_use$ 

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

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

verbose verbose

#### **Details**

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

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

#### Value

giotto object with new subclusters appended to cell metadata

#### See Also

```
doLouvainCluster_multinet and doLouvainCluster_community
```

### **Examples**

```
doLouvainSubCluster(gobject)
```

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

# **Description**

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

# Usage

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

```
gobject giotto object

name name for new clustering result

cluster_column cluster column to subcluster

selected_clusters

only do subclustering on these clusters
```

hvg\_param parameters for calculateHVG

hvg\_min\_perc\_cells

threshold for detection in min percentage of cells

hvg\_mean\_expr\_det

threshold for mean expression level in cells with detection

use\_all\_genes\_as\_hvg

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

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

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

resolution resolution

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

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

verbose verbose

#### Details

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

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

#### Value

giotto object with new subclusters appended to cell metadata

#### See Also

doLouvainCluster\_community

# **Examples**

doLouvainSubCluster\_community(gobject)

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

# Description

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

### Usage

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

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

136 doRandomWalkCluster

nn\_param parameters for parameters for createNearestNetwork

k\_neighbors number of k for createNearestNetwork

gamma gamma omega omega

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

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

verbose verbose

python\_path specify specific path to python if required

### **Details**

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

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

# Value

giotto object with new subclusters appended to cell metadata

#### See Also

doLouvainCluster\_multinet

# **Examples**

doLouvainSubCluster\_multinet(gobject)

 ${\tt doRandomWalkCluster} \qquad doRandomWalkCluster$ 

# Description

Cluster cells using a random walk approach.

doRandomWalkCluster 137

### Usage

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

### **Arguments**

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

# **Details**

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

# Value

giotto object with new clusters appended to cell metadata

# **Examples**

```
doRandomWalkCluster(gobject)
```

138 doSNNCluster

doSNNCluster doSNNCluster

# Description

Cluster cells using a SNN cluster approach.

# Usage

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

# Arguments

gobject giotto object name name for cluster

nn\_network\_to\_use

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

network\_name name of kNN network to use

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

graph.

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

neighbors.

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

to be considered a core points.

borderPoints should borderPoints be assigned to clusters like in DBSCAN?

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

set\_seed set seed

seed\_number number for seed

# **Details**

See sNNclust from dbscan package

### Value

giotto object with new clusters appended to cell metadata

do\_cell\_proximity\_test 139

### **Examples**

```
doSNNCluster(gobject)
```

# Description

Performs a selected differential test on subsets of a matrix

# Usage

### **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, mean_method, offset = 0.1)
```

### **Examples**

```
do_limmatest()
```

do\_page\_permutation

```
\label{local_do_multi_permuttest_random} do\_multi\_permuttest\_random
```

# Description

calculate multiple random values

# Usage

```
do_multi_permuttest_random(
  expr_values,
  select_ind,
  other_ind,
  mean_method,
  offset = 0.1,
  n = 100,
  cores = 2
```

# **Examples**

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

```
do\_page\_permutation do\_page\_permutation
```

# Description

creates permutation for the PAGEEnrich test

# Usage

```
do_page_permutation(gobject, sig_gene, ntimes)
```

# **Examples**

```
do_page_permutation()
```

```
do_permuttest_original
```

 $do\_permuttest\_original$ 

# Description

calculate original values

# Usage

```
do_permuttest_original(
  expr_values,
  select_ind,
  other_ind,
  name = "orig",
  mean_method,
  offset = 0.1
)
```

# **Examples**

```
do_permuttest_original()
```

```
do_permuttest_random do_permuttest_random
```

# Description

calculate random values

Performs permutation test on subsets of a matrix

```
do_permuttest_random(
   expr_values,
   select_ind,
   other_ind,
   name = "perm_1",
   mean_method,
   offset = 0.1
)

do_permuttest(
   expr_values,
   select_ind,
   other_ind,
   n_perm = 1000,
   adjust_method = "fdr",
   mean_method,
```

```
offset = 0.1,
cores = 2
)
```

# **Examples**

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

 ${\tt do\_rank\_permutation} \qquad {\tt do\_rank\_permutation}$ 

# **Description**

creates permutation for the rankEnrich test

# Usage

```
do_rank_permutation(sc_gene, n)
```

# **Examples**

```
do_rank_permutation()
```

```
\begin{tabular}{ll} do\_spatial\_grid\_averaging \\ & do\_spatial\_grid\_averaging \\ \end{tabular}
```

# 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
)
```

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

### Value

matrix with smoothened gene expression values based on spatial grid

### **Examples**

```
do_spatial_grid_averaging(gobject)
```

### **Description**

smooth gene expression over a kNN spatial network

### Usage

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

### **Arguments**

### **Details**

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

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

### Value

matrix with smoothened gene expression values based on kNN spatial network

### **Examples**

```
{\tt do\_spatial\_knn\_smoothing(gobject)}
```

DT\_removeNA

do\_ttest

do\_ttest

# Description

Performs t.test on subsets of a matrix

Performs wilcoxon on subsets of a matrix

# Usage

```
do_ttest(
   expr_values,
   select_ind,
   other_ind,
   adjust_method,
   offset = 0.1
)

do_wilctest(
   expr_values,
   select_ind,
   other_ind,
   adjust_method,
   mean_method,
   offset = 0.1
)
```

# Examples

do\_ttest()
do\_ttest()

DT\_removeNA

DT\_removeNA

# **Description**

set NA values to 0 in a data.table object

```
DT_removeNA(DT)
```

dt\_to\_matrix 145

dt\_to\_matrix

dt\_to\_matrix

## Description

converts data.table to matrix

#### Usage

```
dt_to_matrix(x)
```

### **Examples**

```
dt_to_matrix(x)
```

estimateCellCellDistance

estimate Cell Cell Distance

## Description

estimate average distance between neighboring cells

## Usage

```
estimateCellCellDistance(
  gobject,
  spatial_network_name = "Delaunay_network",
  method = c("mean", "median")
)
```

 $\verb"estimateImageBg"$ 

estimateImageBg

## **Description**

helps to estimate which color is the background color of your plot

## Usage

```
estimateImageBg(mg_object, top_color_range = 1:50)
```

# Arguments

```
mg_object magick image or Giotto image object top_color_range top possible background colors to return
```

146 exportGiottoViewer

#### Value

vector of pixel color frequencies and an associated barplot

## **Examples**

```
estimateImageBg(mg_object)
```

```
evaluate_expr_matrix evaluate_expr_matrix
```

## Description

Evaluate expression matrices.

## Usage

```
evaluate_expr_matrix(inputmatrix, sparse = TRUE, cores = NA)
```

# Arguments

```
inputmatrix inputmatrix to evaluate
```

## **Details**

The inputmatrix can be a matrix, sparse matrix, data.frame, data.table or path to any of these.

## Value

sparse matrix

## **Examples**

```
evaluate_expr_matrix()
```

exportGiottoViewer

exportGiottoViewer

## Description

compute highly variable genes

*exportGiottoViewer* 147

#### Usage

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

### **Arguments**

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

#### **Details**

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

#### Value

writes the necessary output to use in Giotto Viewer

148 exprCellCellcom

#### **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)
```

extended\_gini\_fun 149

```
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

```
extend_vector
```

extend\_vector

### **Description**

extend the range of a vector by a given ratio

#### Usage

```
extend_vector(x, extend_ratio)
```

```
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

## **Examples**

extractNearestNetwork(gobject)

fDataDT

fDataDT

## Description

show gene metadata

## Usage

```
fDataDT(gobject)
```

### **Arguments**

gobject

giotto object

#### Value

data.table with gene metadata

## **Examples**

```
pDataDT(gobject)
```

 $\verb|filterCellProximityGenes||$ 

filterCellProximityGenes

## Description

Filter cell proximity gene scores.

```
filterCellProximityGenes(
  cpgObject,
  min_cells = 4,
  min_cells_expr = 1,
  min_int_cells = 4,
  min_int_cells_expr = 1,
  min_fdr = 0.1,
  min_spat_diff = 0.2,
  min_log2_fc = 0.2,
  min_zscore = 2,
  zscores_column = c("cell_type", "genes"),
  direction = c("both", "up", "down")
)
```

filterCombinations 151

#### **Arguments**

```
cpgObject
                 cell proximity gene score object
                 minimum number of source cell type
min_cells
min_cells_expr minimum expression level for source cell type
min_int_cells
                 minimum number of interacting neighbor cell type
min_int_cells_expr
                 minimum expression level for interacting neighbor cell type
min_fdr
                 minimum adjusted p-value
min_spat_diff
                 minimum absolute spatial expression difference
                 minimum log2 fold-change
min_log2_fc
min_zscore
                 minimum z-score change
zscores_column calculate z-scores over cell types or genes
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.

```
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,
  return_plot = FALSE,
  save_plot = NA,
  save_param = list(),
  default_save_name = "filterCombinations")
```

152 filterCPG

#### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
{\tt expression\_thresholds}
                  all thresholds to consider a gene expressed
gene_det_in_min_cells
                  minimum number of cells that should express a gene to consider that gene fur-
                  ther
min_det_genes_per_cell
                  minimum number of expressed genes per cell to consider that cell further
scale_x_axis
                  ggplot transformation for x-axis (e.g. log2)
x_axis_offset
                  x-axis offset to be used together with the scaling transformation
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
                  return only 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
```

## **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.

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

### Value

list of data.table and ggplot object

### **Examples**

filterCombinations(gobject)

filterCPG filterCPG

## **Description**

Filter cell proximity gene scores.

filterDistributions 153

#### Usage

```
filterCPG(
  cpgObject,
  min_cells = 4,
  min_cells_expr = 1,
  min_int_cells = 4,
  min_int_cells_expr = 1,
  min_fdr = 0.1,
  min_spat_diff = 0.2,
  min_log2_fc = 0.2,
  min_zscore = 2,
  zscores_column = c("cell_type", "genes"),
  direction = c("both", "up", "down")
)
```

## **Arguments**

```
cell proximity gene score object
cpgObject
                 minimum number of source cell type
min_cells
min_cells_expr minimum expression level for source cell type
min_int_cells
                 minimum number of interacting neighbor cell type
min_int_cells_expr
                 minimum expression level for interacting neighbor cell type
min_fdr
                 minimum adjusted p-value
                 minimum absolute spatial expression difference
min_spat_diff
                 minimum log2 fold-change
min_log2_fc
min_zscore
                 minimum z-score change
zscores_column calculate z-scores over cell types or genes
direction
                 differential expression directions to keep
```

### Value

cpgObject that contains the filtered differential gene scores

### **Examples**

```
filterCPG(gobject)
```

 ${\it filter Distributions} \qquad {\it filter Distributions}$ 

## Description

show gene or cell distribution after filtering on expression threshold

154 filterDistributions

#### 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 = NA,
  return_plot = NA,
  save_plot = NA,
 save_param = list(),
 default_save_name = "filterDistributions"
)
```

## **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)
axis_offset
                  offset to be used together with the scaling transformation
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

#### Value

ggplot object

```
filterDistributions(gobject)
```

filterGiotto 155

filterGiotto

filter Giotto

## 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)
```

filter\_network

filter\_network

### **Description**

function to filter a spatial network

## Usage

```
filter_network(networkDT, maximum_distance = NULL, minimum_k = NULL)
```

findCellProximityGenes

findCellProximityGenes

## Description

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

### Usage

```
findCellProximityGenes(
  gobject,
  expression_values = "normalized",
  selected_genes = NULL,
  cluster_column,
  spatial_network_name = "Delaunay_network",
  minimum_unique_cells = 1,
  minimum_unique_int_cells = 1,
  diff_test = c("permutation", "limma", "t.test", "wilcox"),
  mean_method = c("arithmic", "geometric"),
  offset = 0.1,
 adjust_method = c("bonferroni", "BH", "holm", "hochberg", "hommel", "BY", "fdr",
    "none"),
  nr_permutations = 1000,
  exclude_selected_cells_from_test = T,
  do_parallel = TRUE,
  cores = NA
)
```

## **Arguments**

```
gobject giotto object
expression_values
expression values to use
selected_genes subset of selected genes (optional)
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 method to use to calculate the mean mean\_method offset offset value to use when calculating log2 ratio 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 do\_parallel run calculations in parallel with mclapply

#### Details

cores

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

number of cores to use if do\_parallel = TRUE

- 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)

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

## Description

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

#### Usage

```
findCellProximityGenes_per_interaction(
  expr_values,
 cell_metadata,
 annot_spatnetwork,
 sel_int,
 cluster_column = NULL,
 minimum_unique_cells = 1,
 minimum_unique_int_cells = 1,
 exclude_selected_cells_from_test = T,
 diff_test = c("permutation", "limma", "t.test", "wilcox"),
 mean_method = c("arithmic", "geometric"),
 offset = 0.1,
 adjust_method = "bonferroni",
 nr_permutations = 100,
 cores = 1
)
```

#### **Examples**

 ${\tt findCellProximityGenes\_per\_interaction()}$ 

findCPG

findCPG

### **Description**

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

```
findCPG(
  gobject,
  expression_values = "normalized",
  selected_genes = NULL,
  cluster_column,
  spatial_network_name = "Delaunay_network",
  minimum_unique_cells = 1,
  minimum_unique_int_cells = 1,
  diff_test = c("permutation", "limma", "t.test", "wilcox"),
```

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## Arguments

```
gobject
                  giotto object
expression_values
                  expression values to use
selected_genes subset of selected genes (optional)
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
mean_method
                  method to use to calculate the mean
offset
                  offset value to use when calculating log2 ratio
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

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- 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.

## Usage

```
findGiniMarkers(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  group_1 = NULL,
  group_2 = NULL,
  min_expr_gini_score = 0.2,
  min_det_gini_score = 0.2,
  detection_threshold = 0,
  rank_score = 1,
  min_genes = 5
)
```

#### **Arguments**

```
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\_score rank scores for both detection and expression to include

min\_genes minimum number of top genes to return

#### **Details**

Detection of marker genes using the <a href="https://en.wikipedia.org/wiki/Gini\_coefficientgini">https://en.wikipedia.org/wiki/Gini\_coefficientgini</a> coefficient is based on the following steps/principles per gene:

- 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)

findGiniMarkers\_one\_vs\_all

findGiniMarkers\_one\_vs\_all

### **Description**

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

### Usage

```
findGiniMarkers_one_vs_all(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  cluster_column,
  subset_clusters = NULL,
  min_expr_gini_score = 0.5,
  min_det_gini_score = 0.5,
  detection_threshold = 0,
  rank_score = 1,
  min_genes = 4,
  verbose = TRUE
)
```

## **Arguments**

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

### Value

data.table with marker genes

### See Also

findGiniMarkers

```
findGiniMarkers_one_vs_all(gobject)
```

findMarkers 163

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 cluster IDs from cluster_column for pairwise comparison
group_1
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
min_expr_gini_score
                  gini: filter on minimum gini coefficient for expression
min_det_gini_score
                  gini: filter minimum gini coefficient for detection
{\tt detection\_threshold}
                  gini: detection threshold for gene expression
                  gini: rank scores to include
rank_score
                  minimum number of top genes to return (for gini)
min_genes
                  mast: custom name for group_1 clusters
group_1_name
```

```
group_2_name mast: custom name for group_2 clusters

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

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

#### **Details**

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

#### Value

data.table with marker genes

#### See Also

findScranMarkers, findGiniMarkers and findMastMarkers

#### **Examples**

```
findMarkers(gobject)
```

## Description

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

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

findMarkers\_one\_vs\_all 165

### 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)
```

166 findMastMarkers

findMastMarkers

findMastMarkers

### **Description**

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

### Usage

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

## **Arguments**

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
                  group 1 cluster IDs from cluster_column for pairwise comparison
group_1
                  custom name for group_1 clusters
group_1_name
                  group 2 cluster IDs from cluster_column for pairwise comparison
group_2
group_2_name
                  custom name for group_2 clusters
adjust_columns column in pDataDT to adjust for (e.g. detection rate)
                  additional parameters for the zlm function in MAST
. . .
```

### **Details**

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

## Value

data.table with marker genes

```
findMastMarkers(gobject)
```

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

### **Description**

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

## Usage

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

## **Arguments**

```
gobject
                  giotto object
expression_values
                  gene expression values to use
cluster_column clusters to use
subset_clusters
                  selection of clusters to compare
adjust_columns column in pDataDT to adjust for (e.g. detection rate)
                  filter on minimal p-value
pval
logFC
                  filter on logFC
min_genes
                  minimum genes to keep per cluster, overrides pval and logFC
                  be verbose
verbose
                  additional parameters for the zlm function in MAST
```

## Value

data.table with marker genes

## See Also

findMastMarkers

```
findMastMarkers_one_vs_all(gobject)
```

168 findScranMarkers

```
findNetworkNeighbors findNetworkNeighbors
```

#### **Description**

Find the spatial neighbors for a selected group of cells within the selected spatial network.

## Usage

```
findNetworkNeighbors(
  gobject,
  spatial_network_name,
  source_cell_ids = NULL,
  name = "nb_cells"
)
```

## **Arguments**

## Value

data.table

### **Examples**

findNetworkNeighbors(gobject)

 ${\tt findScranMarkers}$ 

findScranMarkers

# Description

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

```
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

### **Examples**

```
findScranMarkers(gobject)
```

## **Description**

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

```
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,
  ...
)
```

find\_grid\_3D

## **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

min\_genes minimum genes to keep per cluster, overrides pval and logFC

verbose be verbose

... additional parameters for the findMarkers function in scran

#### Value

data.table with marker genes

### See Also

findScranMarkers

## **Examples**

findScranMarkers\_one\_vs\_all(gobject)

find\_grid\_2D

 $find\_grid\_2D$ 

# Description

find grid location in 2D

## Usage

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

find\_grid\_3D

find\_grid\_3D

# Description

find grid location in 3D

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

find\_grid\_x

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

# Usage

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

find\_grid\_z

find\_grid\_z

## Description

find grid location on z-axis

## Usage

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

find\_x\_y\_ranges

find\_x\_y\_ranges

# Description

get the extended ranges of x and y

```
find_x_y_ranges(data, extend_ratio)
```

172 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 173

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

sparse expression matrix from 10X

## **Examples**

```
get10Xmatrix(path_to_data)
```

```
getClusterSimilarity
getClusterSimilarity
```

#### **Description**

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

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

174 getDendrogramSplits

#### **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..

## 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
```

getDistinctColors 175

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

verbose be verbose

#### **Details**

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

### Value

data.table object

### **Examples**

getDendrogramSplits(gobject)

getDistinctColors get

getDistinctColors

## Description

Returns a number of distint colors based on the RGB scale

# Usage

```
getDistinctColors(n)
```

## Arguments

n number of colors wanted

### Value

number of distinct colors

get\_distance

getGiottoImage

getGiottoImage

## **Description**

```
get get a giotto image from a giotto object
```

### Usage

```
getGiottoImage(gobject, image_name)
```

## Arguments

gobject

giotto object

 $image\_name$ 

name of giotto image showImageNames

#### Value

a giotto image

## **Examples**

```
getGiottoImage(gobject)
```

```
{\it get\_cross\_section\_coordinates} \\ {\it get\_cross\_section\_coordinates}
```

## Description

get local coordinates within cross section plane

## Usage

```
get_cross_section_coordinates(cell_subset_projection_locations)
```

get\_distance

get\_distance

### **Description**

estimate average distance between neighboring cells with network table as input

```
get_distance(networkDT, method = c("mean", "median"))
```

get\_os 177

get\_os

get\_os

# Description

 $return\ the\ type\ of\ operating\ system,\ see\ https://conjugateprior.org/2015/06/identifying-the-os-from-r/$ 

# Usage

```
get_os()
```

### Value

character osx, linux or windows

```
get\_sectionThickness get\_sectionThickness
```

### **Description**

get section thickness

## Usage

```
get_sectionThickness(
  gobject,
  thickness_unit = c("cell", "natural"),
  slice_thickness = 2,
  spatial_network_name = "Delaunay_network",
  cell_distance_estimate_method = c("mean", "median"),
  plane_equation = NULL
)
```

```
ggplot_save_function ggplot_save_function
```

## Description

Function to automatically save plots to directory of interest

178 ggplot\_save\_function

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

to prevent the common error of specifying dimensions in pixels.

#### 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,
)
```

## **Arguments**

giotto object gobject plot\_object ggplot object to plot save\_dir directory to save to save\_folder folder in save\_dir to save to save\_name name of plot format (e.g. png, tiff, pdf, ...) save\_format show\_saved\_plot load & display the saved plot ncol number of columns number of rows nrow scale scale base\_width width base\_height height base\_aspect\_ratio aspect ratio units units dpi Plot resolution

See Also

limitsize

cowplot::save\_plot

giotto-class 179

### **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
{\tt dimension\_reduction}\ \ {\tt 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
```

giotto\_lapply

giotto\_lapply

## **Description**

```
giotto_lapply
```

```
giotto_lapply(X, cores = NA, fun, ...)
```

 $heatmSpatialCorGenes \quad \textit{heatmSpatialCorGenes}$ 

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

### **Arguments**

```
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
. . .
```

hyperGeometricEnrich 181

#### 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

#### **Examples**

```
hyperGeometricEnrich(gobject)
```

insertCrossSectionGenePlot3D

insertCrossSectionGenePlot3D

### **Description**

Visualize cells and gene expression in a virtual cross section according to spatial coordinates

```
insertCrossSectionGenePlot3D(
 gobject,
 crossSection_obj = NULL,
 name = NULL,
 spatial_network_name = "Delaunay_network",
 mesh_grid_color = "#1f77b4",
 mesh_grid_width = 3,
 mesh_grid_style = "dot",
 sdimx = "sdimx",
 sdimy = "sdimy",
  sdimz = "sdimz",
 expression_values = c("normalized", "scaled", "custom"),
 genes,
 show_network = F,
 network_color = NULL,
 edge_alpha = NULL,
  show\_grid = F,
 cluster_column = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = F,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 genes_high_color = NULL,
 genes_mid_color = "white";
 genes_low_color = "darkblue",
  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_with_cross_section"
)
```

#### **Arguments**

```
giotto object
gobject
                  name of virtual cross section to use
spatial_network_name
                  name of spatial network to use
mesh_grid_color
                  color for the meshgrid lines
mesh_grid_width
                  width for the meshgrid lines
mesh_grid_style
                  style for the meshgrid lines
sdimx
                  x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
                  z-axis dimension name (default = 'sdimy')
sdimz
expression_values
                  gene expression values to use
                  genes to show
genes
                  show underlying spatial network
show_network
network_color
                  color of spatial network
show_grid
                  show spatial grid
genes_high_color
                  color represents high gene expression
genes_mid_color
                  color represents middle gene expression
genes_low_color
                  color represents low gene expression
spatial_grid_name
                  name of spatial grid to use
                  size of point (cell)
point_size
                  show legend
show_legend
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
grid_color
                  color of spatial grid
                  expression midpoint
midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  parameters for cowplot::save_plot()
```

#### **Details**

Description of parameters.

#### Value

ggplot

#### **Examples**

insertCrossSectionGenePlot3D(gobject)

```
insertCrossSectionSpatPlot3D
```

insertCrossSectionSpatPlot3D

## **Description**

Visualize the meshgrid lines of cross section together with cells

```
insertCrossSectionSpatPlot3D(
 gobject,
 crossSection_obj = NULL,
 name = NULL,
 spatial_network_name = "Delaunay_network",
 mesh_grid_color = "#1f77b4",
 mesh_grid_width = 3,
 mesh_grid_style = "dot",
 sdimx = "sdimx",
 sdimy = "sdimy",
  sdimz = "sdimz",
 point_size = 2,
 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,
 show\_grid = F,
 grid_color = NULL,
  spatial_grid_name = "spatial_grid",
  title = "",
  show_legend = T,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
```

```
x_ticks = NULL,
      y_ticks = NULL,
      z_ticks = NULL,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spat3D_with_cross_section"
Arguments
    gobject
                      giotto object
                      name of virtual cross section to use
    name
    spatial_network_name
                      name of spatial network to use
    mesh_grid_color
                      color for the meshgrid lines
    mesh_grid_width
                      width for the meshgrid lines
    mesh_grid_style
                      style for the meshgrid lines
    sdimx
                      x-axis dimension name (default = 'sdimx')
    sdimy
                      y-axis dimension name (default = 'sdimy')
                      z-axis dimension name (default = 'sdimy')
    sdimz
                      size of point (cell)
    point_size
    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
    show_other_cells
                      display not selected cells
    other_cell_color
                      color of not selected cells
    other_point_size
                      point size of not selected cells
                      color of spatial network
    network_color
                      show spatial grid
    show_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
                      customize the scale of the plot
    custom_ratio
```

set the number of ticks on the x-axis

x\_ticks

jackstrawPlot

```
y_ticks set the number of ticks on the y-axis

z_ticks set the number of ticks on the z-axis

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

insertCrossSectionSpatPlot3D(gobject)

jackstrawPlot jackstrawPlot

# Description

identify significant prinicipal components (PCs)

```
jackstrawPlot(
 gobject,
  expression_values = c("normalized", "scaled", "custom"),
 reduction = c("cells", "genes"),
 genes_to_use = NULL,
 center = FALSE,
  scale_unit = FALSE,
 ncp = 20,
 ylim = c(0, 1),
  iter = 10,
  threshold = 0.01,
  verbose = T,
  show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "jackstrawPlot"
)
```

kmeans\_binarize 187

## **Arguments**

gobject giotto object

expression\_values

expression values to use

reduction cells or genes

genes\_to\_use subset of genes to use for PCA

center center data before PCA scale\_unit scale features before PCA

ncp number of principal components to calculate

ylim y-axis limits on jackstraw plot

iter number of interations for jackstraw

threshold p-value threshold to call a PC significant verbose show progress of jackstraw method

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**

The Jackstraw method uses the permutationPA function. By systematically permuting genes it identifies robust, and thus significant, PCs.

## Value

ggplot object for jackstraw method

## **Examples**

jackstrawPlot(gobject)

kmeans\_binarize

kmeans\_binarize

# **Description**

create binarized scores from a vector using kmeans

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

188 loadHMRF

libNorm\_giotto

libNorm\_giotto

## **Description**

libNorm\_giotto

## Usage

```
libNorm_giotto(mymatrix, scalefactor)
```

loadHMRF

loadHMRF

# Description

load previous HMRF

# Usage

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

# Arguments

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

# **Details**

Description of HMRF parameters ...

# Value

reloads a previous ran HMRF from doHRMF

```
loadHMRF(gobject)
```

logNorm\_giotto 189

logNorm\_giotto

logNorm\_giotto

## **Description**

logNorm\_giotto

## Usage

logNorm\_giotto(mymatrix, base, offset)

makeSignMatrixPAGE

make Sign Matrix PAGE

# Description

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

makeSignMatrixRank

makeSignMatrixRank

## **Description**

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

## Usage

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

#### **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 considered
```

## 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 = "Delaunay_network",
  cluster_column,
  number_of_simulations = 100
)
```

```
make_simulated_network(gobject)
```

mean\_expr\_det\_test 191

# Description

```
mean_expr_det_test
```

# Usage

```
mean_expr_det_test(mymatrix, detection_threshold = 1)
```

 $mean\_giotto$ 

mean\_giotto

# Description

```
mean_giotto
```

## Usage

```
mean_giotto(x, ...)
```

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

192 mygini\_fun

#### **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
new_cluster_name
                  new name for merged clusters
min_cor_score min correlation score to merge pairwise clusters
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

my\_arowMeans 193

my\_arowMeans

 $my\_arowMeans$ 

# Description

arithmic rowMeans that works for a single column

# Usage

```
my_arowMeans(x)
```

# **Examples**

```
my_arowMeans(x)
```

my\_growMeans

 $my\_growMeans$ 

# Description

geometric rowMeans that works for a single column

# Usage

```
my\_growMeans(x, offset = 0.1)
```

# **Examples**

my\_growMeans(x)

my\_rowMeans

my\_rowMeans

# Description

arithmic or geometric rowMeans that works for a single column

# Usage

```
my_rowMeans(x, method = c("arithmic", "geometric"), offset = 0.1)
```

```
my_rowMeans(x)
```

node\_clusters

nnDT\_to\_kNN

 $nnDT\_to\_kNN$ 

# Description

Convert a nearest network data.table to a kNN object

# Usage

```
nnDT_to_kNN(nnDT)
```

# **Arguments**

nnDT

nearest neighbor network in data.table format

## Value

kNN object

node\_clusters

node\_clusters

# Description

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

# Usage

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

# Arguments

hclus\_obj hclus object verbose be verbose

# Value

list of splitted dendrogram nodes from high to low node height

```
node_clusters(hclus_obj)
```

normalizeGiotto 195

normalizeGiotto	normalizeGiotto
-----------------	-----------------

## **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.

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- 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)
```

**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
```

pca\_giotto 197

#### **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

#### **Examples**

PAGEEnrich(gobject)

pca\_giotto

pca\_giotto

# Description

performs PCA based on Rfast

# Usage

```
pca_giotto(mymatrix, center = T, scale = T, k = 50)
```

## **Arguments**

mymatrix matrix or object that can be converted to matrix

center center data
scale scale features

k number of principal components to calculate

#### Value

list of eigenvalues, eigenvectors and pca coordinates

198 plotCCcomDotplot

pDataDT

pDataDT

## **Description**

show cell metadata

## Usage

```
pDataDT(gobject)
```

#### **Arguments**

gobject

giotto object

#### Value

data.table with cell metadata

## **Examples**

```
pDataDT(gobject)
```

plotCCcomDotplot

plotCCcomDotplot

# Description

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

```
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"
)
```

plotCCcomHeatmap 199

#### **Arguments**

```
gobject
                  giotto object
                  communinication scores from exprCellCellcom or spatCellCellcom
comScores
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
                  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_plot
                  show plots
return_plot
                  return plotting 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
show
                  values to show on heatmap
```

#### Value

ggplot

## **Examples**

```
plotCCcomDotplot(CPGscores)
```

plotCCcomHeatmap plotCCcomHeatmap

## **Description**

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

```
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")
```

# Arguments

gobject giotto object comScores  $communinication\ scores\ from\ expr\cell\cellcom\ or\ spat\cell\cellcom$ selected\_LR selected ligand-receptor combinations 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 show values to show on heatmap cor\_method correlation method used for clustering agglomeration method used by hclust  $aggl_method$ 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

ggplot

# **Examples**

```
plotCCcomHeatmap(CPGscores)
```

```
plot Cell Proximity Genes \\ plot Cell Proximity Genes
```

# 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 = 4,
  min_cells_expr = 1,
  min_int_cells = 4,
  min_int_cells_expr = 1,
  min_fdr = 0.1,
  min_spat_diff = 0.2,
  min_log2_fc = 0.2,
  min_zscore = 2,
  zscores_column = c("cell_type", "genes"),
  direction = c("both", "up", "down"),
  cell_color_code = NULL,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCellProximityGenes"
```

### **Arguments**

```
gobject
                  giotto object
cpgObject
                  cell proximity gene score object
method
                  plotting method to use
min_cells
                  minimum number of source cell type
min_cells_expr minimum expression level for source cell type
min_int_cells
                  minimum number of interacting neighbor cell type
min_int_cells_expr
                  minimum expression level for interacting neighbor cell type
min_fdr
                  minimum adjusted p-value
min_spat_diff
                  minimum absolute spatial expression difference
                  minimum log2 fold-change
min_log2_fc
min_zscore
                  minimum z-score change
zscores_column calculate z-scores over cell types or genes
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

202 plotCombineCCcom

#### Value

plot

## **Examples**

```
plotCellProximityGenes(CPGscores)
```

plotCombineCCcom

plotCombineCCcom

### **Description**

Create visualization for combined (pairwise) cell proximity gene scores

## Usage

```
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**

```
gobject
                  giotto object
{\sf 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
{\tt 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

## **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**

```
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**

```
plotCombineCellCellCommunication(CPGscores)
```

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

## **Description**

Create visualization for combined (pairwise) cell proximity gene scores

```
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"
)
```

#### **Arguments**

```
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
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

```
plotCombineCellProximityGenes(CPGscores)
```

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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"
)
```

### **Arguments**

```
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
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

plotCPG 207

#### Value

ggplot

#### **Examples**

```
plotCombineCPG(CPGscores)
```

plotCPG

plotCPG

## **Description**

Create visualization for cell proximity gene scores

## Usage

```
plotCPG(
  gobject,
  cpgObject,
  method = c("volcano", "cell_barplot", "cell-cell", "cell_sankey", "heatmap",
    "dotplot"),
  min_cells = 5,
  min_cells_expr = 1,
  min_int_cells = 3,
  min_int_cells_expr = 1,
  min_fdr = 0.05,
  min_spat_diff = 0.2,
  min_log2_fc = 0.2,
  min_zscore = 2,
  zscores_column = c("cell_type", "genes"),
  direction = c("both", "up", "down"),
  cell_color_code = NULL,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotCPG"
)
```

# Arguments

```
gobject giotto object
cell proximity gene score object
method plotting method to use
min_cells minimum number of source cell type
min_cells_expr minimum expression level for source cell type
min_int_cells minimum number of interacting neighbor cell type
min_int_cells_expr
minimum expression level for interacting neighbor cell type
```

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min\_fdr minimum adjusted p-value

min\_spat\_diff minimum absolute spatial expression difference

min\_log2\_fc minimum log2 fold-change min\_zscore minimum z-score change

zscores\_column calculate z-scores over cell types or genes 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]

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

plot

# **Examples**

plotCPG(CPGscores)

plotGiottoImage plotGiottoImage

# **Description**

get plot a giotto image from a giotto object

## Usage

```
plotGiottoImage(gobject, image_name)
```

# **Arguments**

gobject giotto object

image\_name name of giotto image showImageNames

# Value

plot

```
\verb|plotGiottoImage(gobject)|
```

plotHeatmap 209

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("correlation", "custom"),
  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"
)
```

# 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_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
save_param
                  list of saving parameters, see showSaveParameters
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)

plotICG 211

plotICG plotICG

## **Description**

Create barplot to visualize interaction changed genes

## Usage

```
plotICG(
   gobject,
   cpgObject,
   source_type,
   source_markers,
   ICG_genes,
   cell_color_code = NULL,
   show_plot = NA,
   return_plot = NA,
   save_plot = NA,
   save_param = list(),
   default_save_name = "plotICG"
)
```

# Arguments

```
gobject
                  giotto object
                  cell proximity gene score object
cpgObject
source_type
                  cell type of the source cell
source_markers markers for the source cell type
ICG_genes
                  named character vector of ICG genes
cell_color_code
                  cell color code for the interacting cell types
{\sf 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

plot

```
plotICG(CPGscores)
```

```
plotInteraction {\tt Changed Genes} \\ plotInteraction {\tt Changed Genes} \\
```

# Description

Create barplot to visualize interaction changed genes

# Usage

```
plotInteractionChangedGenes(
   gobject,
   cpgObject,
   source_type,
   source_markers,
   ICG_genes,
   cell_color_code = NULL,
   show_plot = NA,
   return_plot = NA,
   save_plot = NA,
   save_param = list(),
   default_save_name = "plotInteractionChangedGenes")
```

# **Arguments**

```
gobject
                  giotto object
cpgObject
                  cell proximity gene score object
                  cell type of the source cell
source_type
source_markers markers for the source cell type
ICG_genes
                  named character vector of ICG genes
cell_color_code
                  cell color code for the interacting cell types
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

```
plotInteractionChangedGenes(CPGscores)
```

plotly\_axis\_scale\_2D 213

```
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

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

214 plotly\_grid

# **Arguments**

#### Value

edges in spatial grid as data.table()

## **Examples**

```
plotly_axis_scale_3D(gobject)
```

plotly\_grid

plotly\_grid

# Description

provide grid segment to draw in plot\_ly()

# Usage

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

# Arguments

```
spatial_grid spatial_grid in giotto object
```

# Value

edges in spatial grid as data.table()

```
plotly_grid(gobject)
```

plotly\_network 215

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.

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

custom\_gene\_order

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

custom gene order (default = NULL)

plotMetaDataHeatmap 217

#### **Details**

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

### Value

ggplot or data.table

#### See Also

plotMetaDataHeatmap for gene expression instead of numeric cell annotation data.

#### **Examples**

```
plotMetaDataCellsHeatmap(gobject)
```

```
plotMetaDataHeatmap
```

### **Description**

Creates heatmap for genes within aggregated clusters.

```
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",
  gradient_color = c("blue", "white", "red"),
  gradient_midpoint = 0,
  gradient_limits = NULL,
  x_{text_size} = 10,
  x_{text_angle} = 45,
  y_{text_size} = 10,
  strip_text_size = 8,
  show_plot = NA,
```

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

```
gobject
                 giotto object
expression_values
                 expression values to use
metadata_cols annotation columns found in pDataDT(gobject)
selected_genes subset of genes to use
first_meta_col if more than 1 metadata column, select the x-axis factor
second_meta_col
                 if more than 1 metadata column, select the facetting factor
show_values
                 which values to show on heatmap
custom_cluster_order
                 custom cluster order (default = NULL)
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
gradient_color vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                 vector with lower and upper limits
                 size of x-axis text
x_text_size
x_text_angle
                 angle of x-axis text
y_text_size
                 size of y-axis text
strip_text_size
                 size of strip text
show_plot
                 show plot
return_plot
                 return ggplot object
save_plot
                 directly save the plot [boolean]
                 list of saving parameters, see showSaveParameters
save_param
default_save_name
                 default save name
```

plotPCA 219

#### **Details**

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

#### Value

ggplot or data.table

#### See Also

plotMetaDataCellsHeatmap for numeric cell annotation instead of gene expression.

# **Examples**

```
plotMetaDataHeatmap(gobject)
```

plotPCA

plotPCA

### **Description**

Short wrapper for PCA visualization

# Usage

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

```
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
groub_by
                 create multiple plots based on cell annotation column
group_by_subset
                 subset the group_by factor column
                 dimension to use on x-axis
dim1_to_use
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)
```

220 plotPCA

color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_shape point with border or not (border or no border) point\_size size of point (cell) point\_alpha transparancy of point 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 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

plotPCA\_2D 221

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

plotPCA\_2D

show\_NN\_network show underlying NN network nn\_network\_to\_use type of NN network to use (kNN vs sNN) name of NN network to use, if show\_NN\_network = TRUE network\_name cell\_color color for cells (see details) color\_as\_factor convert color column to factor cell\_color\_code named vector with colors cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size size of not selected cells show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels edge\_alpha column to use for alpha of the edges point\_shape point with border or not (border or no\_border) point\_size size of point (cell) point\_alpha transparancy of point 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

plotPCA\_3D 223

```
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 cowplot param: relative height cow\_rel\_h cowplot param: relative width cow\_rel\_w cow\_align cowplot param: how to align show plot show\_plot return\_plot return ggplot object save\_plot directly save the plot [boolean] list of saving parameters from all\_plots\_save\_function save\_param

# **Details**

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

### Value

ggplot

# **Examples**

```
plotPCA_2D(gobject)
```

plotPCA\_3D plotPCA\_3D

# Description

Visualize cells according to 3D PCA dimension reduction

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

224 plotPCA\_3D

#### **Arguments**

```
gobject
                  giotto object
dim_reduction_name
                  pca dimension reduction name
default_save_name
                  default save name for saving, ideally change save_name in save_param
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
dim3_to_use
                  dimension to use on z-axis
show_NN_network
                  show underlying NN network
nn_network_to_use
                  type of NN network to use (kNN vs sNN)
                  name of NN network to use, if show_NN_network = TRUE
network_name
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select 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
save_param
                  list of saving parameters from all_plots_save_function
```

#### **Details**

Description of parameters.

plotRankSpatvsExpr 225

#### Value

plotly

#### **Examples**

```
plotPCA_3D(gobject)
```

plotRankSpatvsExpr

plotRankSpatvsExpr

#### **Description**

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

# 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"
)
```

```
giotto object
gobject
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 of colors
midpoint
                  size ranges of dotplot
size_range
xlims
                  x-limits, numerical vector of 2
ylims
                  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
```

226 plotRecovery

#### Value

ggplot

### **Examples**

```
plotRankSpatvsExpr(CPGscores)
```

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_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

plotRecovery\_sub 227

#### Value

ggplot

### **Examples**

```
plotRecovery(CPGscores)
```

```
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**

```
{\tt plotRecovery\_sub(CPGscores)}
```

```
plotStatDelaunayNetwork
```

plot Stat Delaunay Network

### **Description**

Plots network statistics for a Delaunay network..

```
plotStatDelaunayNetwork(
  gobject,
  method = c("deldir", "delaunayn_geometry", "RTriangle"),
  dimensions = "all",
  maximum_distance = "auto",
  minimum_k = 0,
  options = "Pp",
  Y = TRUE,
  j = TRUE,
  S = 0,
  show_plot = NA,
```

```
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "plotStatDelaunayNetwork",
...
)
```

#### **Arguments**

gobject giotto object which spatial dimensions to use (maximum 2 dimensions) dimensions maximum\_distance distance cuttof for Delaunay neighbors to consider minimum k minimum neighbours if maximum distance != NULL (geometry) String containing extra control options for the underlying Qhull options command; see the Qhull documentation (../doc/qhull/html/qdelaun.html) for the available options. (default = 'Pp', do not report precision problems) (RTriangle) If TRUE prohibits the insertion of Steiner points on the mesh bound-Υ (RTriangle) If TRUE jettisons vertices that are not part of the final triangulation j 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] 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 Other parameters of the triangulate function . . . name for spatial network (default = 'delaunay\_network') name

### **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**

```
plotStatDelaunayNetwork(gobject)
```

plotTSNE 229

*plotTSNE* plotTSNE

### **Description**

Short wrapper for tSNE visualization

#### Usage

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

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

plotTSNE

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

#### **Details**

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

### Value

ggplot

# **Examples**

```
plotTSNE(gobject)
```

plotTSNE\_2D 231

plotTSNE\_2D

plotTSNE\_2D

#### **Description**

Short wrapper for tSNE visualization

# Usage

```
plotTSNE_2D(
  gobject,
  dim_reduction_name = "tsne",
  default_save_name = "tSNE_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)
network_name
                 name of NN network to use, if show_NN_network = TRUE
                 color for cells (see details)
cell_color
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                 select subset of cells/clusters based on cell_color parameter
```

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```
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
                  point with border or not (border or no_border)
point_shape
                  size of point (cell)
point_size
point_alpha
                  transparancy of point
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
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
```

### **Details**

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

plotTSNE\_3D 233

#### Value

ggplot

#### **Examples**

```
plotTSNE_2D(gobject)
```

plotTSNE\_3D

plotTSNE 3D

### **Description**

Visualize cells according to dimension reduction coordinates

#### Usage

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

```
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
```

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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
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**

plotTSNE\_3D(gobject)

plotUMAP plotUMAP

# Description

Short wrapper for UMAP visualization

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

plotUMAP 235

#### **Arguments**

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

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point\_size size of point (cell)

point\_alpha transparancy of point

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\_hcowplot param: relative heightcow\_rel\_wcowplot param: relative width

cow\_align cowplot param: how to align

 $\verb|show_plot| & show plot|$ 

return\_plot return ggplot object

save\_plot directly save the plot [boolean]

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

### **Details**

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

#### Value

ggplot

### **Examples**

plotUMAP(gobject)

plotUMAP\_2D 237

plotUMAP\_2D

plotUMAP\_2D

#### **Description**

Short wrapper for UMAP visualization

# Usage

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

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

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```
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
                  point with border or not (border or no_border)
point_shape
                  size of point (cell)
point_size
point_alpha
                  transparancy of point
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
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
```

### **Details**

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

plotUMAP\_3D 239

#### Value

ggplot

#### **Examples**

```
plotUMAP_2D(gobject)
```

plotUMAP\_3D

plotUMAP\_3D

# Description

Visualize cells according to dimension reduction coordinates

# Usage

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

```
gobject
                  giotto object
\operatorname{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
                  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
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
```

```
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
edge_alpha
                  column to use for alpha of the edges
point_size
                  size of point (cell)
show_legend
                  show legend
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
```

#### **Details**

Description of parameters.

### Value

plotly

### **Examples**

```
plotUMAP_3D(gobject)
```

## **Description**

Visualize cells in network layer according to dimension reduction coordinates

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

#### **Arguments**

#### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
plot_network_layer_ggplot(gobject)
```

### Description

Visualize cells in point layer according to dimension reduction coordinates

```
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_alpha = 1,
  point_border_col = "black",
  point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
```

```
label_size = 4,
      label_fontface = "bold",
      edge_alpha = NULL,
      show_other_cells = T,
      other_cell_color = "lightgrey",
      other_point_size = 0.5,
      show_legend = T
Arguments
    annotated_DT_selected
                      annotated data.table of selected cells
    annotated_DT_other
                      annotated data.table of not selected cells
    cell_color
                      color for cells (see details)
    color_as_factor
                      convert color column to factor
    cell_color_code
                      named vector with colors
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
    select_cells
                      select subset of cells based on cell IDs
    point_size
                      size of point (cell)
                      transparancy of point
    point_alpha
    point_border_col
                      color of border around points
    point_border_stroke
                      stroke size of border around points
    show_cluster_center
                      plot center of selected clusters
    show_center_label
                      plot label of selected clusters
    center_point_size
                      size of center points
    label_size
                      size of labels
    label_fontface font of labels
    edge_alpha
                      column to use for alpha of the edges
    show_other_cells
                      display not selected cells
    other_cell_color
                      color of not selected cells
```

```
other_point_size
size of not selected cells
show_legend
show legend
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

```
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,
  point_alpha = 1,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = 4,
  label_fontface = "bold",
  edge_alpha = NULL,
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 0.5,
  show_legend = T
)
```

#### **Arguments**

```
annotated_DT_selected
                  annotated data.table of selected cells
annotated_DT_other
                  annotated data.table of not selected cells
                  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
                  size of point (cell)
point_size
point_alpha
                  transparancy of point
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
gobject
                  giotto object
```

# **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
{\tt plot\_point\_layer\_ggplot\_noFILL(gobject)}
```

# Description

create background image in ggplot

# Usage

```
plot_spat_image_layer_ggplot(
   ggplot,
   gobject,
   gimage,
   sdimx = NULL,
   sdimy = NULL
)
```

# Arguments

```
gobject giotto object
gimage a giotto image
sdimx x-axis dimension name (default = 'sdimx')
sdimy y-axis dimension name (default = 'sdimy')
```

#### Value

ggplot

# **Examples**

```
plot_spat_image_layer_ggplot(gobject)
```

# Description

creat ggplot point layer for spatial coordinates

#### Usage

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

```
sdimx
                 x-axis dimension name (default = 'sdimx')
                 y-axis dimension name (default = 'sdimy')
sdimy
cell_locations_metadata_selected
                 annotated location from selected cells
cell_locations_metadata_other
                 annotated location from non-selected cells
cell_color
                 color for cells (see details)
color_as_factor
                 convert color column to factor
cell_color_code
                 named vector with colors
cell_color_gradient
                 vector with 3 colors for numeric data
gradient_midpoint
                 midpoint for color gradient
gradient_limits
                 vector with lower and upper limits
```

```
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
                  size of point (cell)
point_size
point_alpha
                  transparancy of point
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
gobject
                  giotto object
```

### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
plot_spat_point_layer_ggplot(gobject)
```

# Description

creat ggplot point layer for spatial coordinates without borders

#### 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,
  point_alpha = 1,
  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
)
```

```
sdimx
                  x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
cell_locations_metadata_selected
                  annotated location from selected cells
cell_locations_metadata_other
                  annotated location from non-selected cells
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell color parameter
                  select subset of cells based on cell IDs
select_cells
```

```
size of point (cell)
point_size
                  transparancy of point
point_alpha
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)
```

### **Description**

creat ggplot point layer for spatial coordinates without borders

```
plot_spat_voronoi_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_alpha = 1,
  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,
 background_color = "white",
  vor_border_color = "white",
 vor_max_radius = 200,
 vor_alpha = 1,
  show_legend = TRUE
)
```

```
sdimx
                  x-axis dimension name (default = 'sdimx')
                  y-axis dimension name (default = 'sdimy')
sdimy
cell_locations_metadata_selected
                  annotated location from selected cells
{\tt cell\_locations\_metadata\_other}
                  annotated location from non-selected cells
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
point_size
                  size of point (cell)
point_alpha
                  transparancy of point
show_cluster_center
                  plot center of selected clusters
```

print.giotto 251

```
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

background\_color

background color

vor\_border\_color

borde colorr of voronoi plot

vor\_max\_radius maximum radius for voronoi 'cells' vor\_alpha transparancy of voronoi 'cells'

show\_legend show legend gobject giotto object

### **Details**

Description of parameters.

### Value

ggplot

# **Examples**

```
plot_spat_voronoi_layer_ggplot(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 252 rankEnrich

projection\_fun

projection\_fun

# Description

project a point onto a plane

### Usage

```
projection_fun(point_to_project, plane_point, plane_norm)
```

rankEnrich

rankEnrich

### **Description**

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

# Usage

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

rankSpatialCorGroups

#### **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.

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

#### Value

data.table with enrichment results

#### See Also

```
makeSignMatrixRank
```

### **Examples**

```
rankEnrich(gobject)
```

```
rankSpatialCorGroups rankSpatialCorGroups
```

# Description

Rank spatial correlated clusters according to correlation structure

# Usage

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

```
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
```

254 readExprMatrix

```
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

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

### **Examples**

```
rankSpatialCorGroups(gobject)
```

rank\_binarize

rank\_binarize

### **Description**

create binarized scores from a vector using arbitrary rank

### Usage

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

readExprMatrix

readExprMatrix

## Description

Function to read an expression matrix into a sparse matrix.

#### Usage

```
readExprMatrix(path, cores = NA, transpose = FALSE)
```

# Arguments

path path to the expression matrix

cores number of cores to use

transpose transpose matrix

### **Details**

The expression matrix needs to have both unique column names and row names

#### Value

sparse matrix

## **Examples**

```
readExprMatrix()
```

readGiottoInstructions 255

```
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()

read\_crossSection

 $read\_crossSection$ 

# Description

read a cross section object from a giotto object

```
read_crossSection(gobject, name = NULL, spatial_network_name = NULL)
```

256 removeGeneAnnotation

removeCellAnnotation removeCellAnnotation

## Description

removes cell annotation of giotto object

### Usage

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

#### **Arguments**

gobject giotto object

columns names of columns to remove

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

#### **Details**

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

### Value

giotto object

## **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

giotto object with replaces instructions

# **Examples**

 ${\tt replaceGiottoInstructions()}$ 

```
reshape_to_data_point reshape_to_data_point
```

## Description

reshape a mesh grid line object to data point matrix

```
reshape_to_data_point(mesh_grid_obj)
```

258 rowSums\_giotto

# Description

reshape a data point matrix to a mesh grid line object

# Usage

```
reshape_to_mesh_grid_obj(data_points, mesh_grid_n)
```

rowMeans\_giotto

rowMeans\_giotto

# Description

 $row Means\_giot to$ 

## Usage

```
rowMeans_giotto(mymatrix)
```

 ${\tt rowSums\_giotto}$ 

rowSums\_giotto

# Description

 $rowSums\_giotto$ 

```
rowSums_giotto(mymatrix)
```

runPCA 259

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 = "hvg",
  return_gobject = TRUE,
  center = F,
  scale_unit = F,
  ncp = 100,
  method = c("irlba", "factominer"),
  rev = FALSE,
  verbose = TRUE,
  ...
)
```

### **Arguments**

```
gobject
                  giotto object
expression_values
                  expression values to use
                  cells or genes
reduction
                  arbitrary name for PCA run
name
                  subset of genes to use for PCA
genes_to_use
return_gobject boolean: return giotto object (default = TRUE)
                  center data first (default = FALSE)
center
scale_unit
                  scale features before PCA
                  number of principal components to calculate
ncp
                  which implementation to use
method
                  do a reverse PCA
rev
                  verbosity of the function
verbose
                  additional parameters for PCA (see details)
. . .
```

#### **Details**

See prcomp\_irlba and PCA for more information about other parameters.

- genes\_to\_use = NULL: will use all genes from the selected matrix
- genes\_to\_use = <hvg name>: can be used to select a column name of highly variable genes, created by (see calculateHVG)
- genes\_to\_use = c('geneA', 'geneB', ...): will use all manually provided genes

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#### Value

giotto object with updated PCA dimension recuction

### **Examples**

```
# 1. create giotto object
expr_path = system.file("extdata", "seqfish_field_expr.txt", package = 'Giotto')
loc_path = system.file("extdata", "seqfish_field_locs.txt", package = 'Giotto')
VC_small <- createGiottoObject(raw_exprs = expr_path, spatial_locs = loc_path)
# 2. normalize giotto
VC_small <- normalizeGiotto(gobject = VC_small, scalefactor = 6000)
VC_small <- addStatistics(gobject = VC_small)
# 3. dimension reduction
VC_small <- calculateHVG(gobject = VC_small)
VC_small <- runPCA(gobject = VC_small)
plotPCA(VC_small)</pre>
```

runPCA\_factominer

runPCA\_factominer

### **Description**

performs PCA based on the factominer package

### Usage

```
runPCA_factominer(x, ncp = 100, scale = TRUE, rev = FALSE, ...)
```

#### **Arguments**

x matrix or object that can be converted to matrix
 ncp number of principal components to calculate
 scale scale features

rev reverse PCA

### Value

list of eigenvalues, loadings and pca coordinates

```
runPCA_prcomp_irlba
```

### **Description**

performs PCA based on the irlba package

### Usage

```
runPCA_prcomp_irlba(
    x,
    ncp = 100,
    center = TRUE,
    scale = TRUE,
    rev = FALSE,
    ...
)
```

## **Arguments**

x matrix or object that can be converted to matrix
ncp number of principal components to calculate
center center data
scale scale features
rev reverse PCA

### Value

list of eigenvalues, loadings and pca coordinates

runtSNE runtSNE

## **Description**

run tSNE

```
runtSNE(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  dim_reduction_to_use = "pca",
  dim_reduction_name = "pca",
  dimensions_to_use = 1:10,
  name = "tsne",
  genes_to_use = NULL,
```

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```
return_gobject = TRUE,
dims = 2,
perplexity = 30,
theta = 0.5,
do_PCA_first = F,
set_seed = T,
seed_number = 1234,
verbose = TRUE,
...
)
```

#### **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 tSNE run
name
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
                  tSNE param: perplexity
perplexity
theta
                  tSNE param: theta
                  tSNE param: do PCA before tSNE (default = FALSE)
do_PCA_first
set_seed
                  use of seed
seed_number
                  seed number to use
verbose
                  verbosity of the function
                  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
- If dim\_reduction\_to\_use = NULL, genes\_to\_use can be used to select a column name of highly variable genes (see calculateHVG) or simply provide a vector of genes
- multiple tSNE results can be stored by changing the name of the analysis

#### Value

giotto object with updated tSNE dimension recuction

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#### **Examples**

```
# 1. create giotto object
expr_path = system.file("extdata", "seqfish_field_expr.txt", package = 'Giotto')
loc_path = system.file("extdata", "seqfish_field_locs.txt", package = 'Giotto')
VC_small <- createGiottoObject(raw_exprs = expr_path, spatial_locs = loc_path)
# 2. normalize giotto
VC_small <- normalizeGiotto(gobject = VC_small, scalefactor = 6000)
VC_small <- addStatistics(gobject = VC_small)
# 3. dimension reduction
VC_small <- calculateHVG(gobject = VC_small)
VC_small <- runPCA(gobject = VC_small)
VC_small <- runTSNE(VC_small, dimensions_to_use = 1:5, n_threads = 2)
plotTSNE(gobject = VC_small)</pre>
```

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,
 verbose = T,
)
```

#### **Arguments**

gobject giotto object

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```
expression_values
                 expression values to use
reduction
                 cells or genes
dim_reduction_to_use
                 use another dimension reduction set as input
dim reduction name
                 name of dimension reduction set to use
dimensions_to_use
                 number of dimensions to use as input
name
                 arbitrary name for UMAP run
                 if dim_reduction_to_use = NULL, which genes to use
genes_to_use
return_gobject boolean: return giotto object (default = TRUE)
                 UMAP param: number of neighbors
n_neighbors
n_components
                 UMAP param: number of components
n_epochs
                 UMAP param: number of epochs
min_dist
                 UMAP param: minimum distance
n_threads
                 UMAP param: threads to use
spread
                 UMAP param: spread
                 use of seed
set\_seed
seed_number
                 seed number to use
verbose
                 verbosity of function
                 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
- If dim\_reduction\_to\_use = NULL, genes\_to\_use can be used to select a column name of highly variable genes (see calculateHVG) or simply provide a vector of genes
- multiple UMAP results can be stored by changing the *name* of the analysis

### Value

giotto object with updated UMAP dimension recuction

### **Examples**

```
# 1. create giotto object
expr_path = system.file("extdata", "seqfish_field_expr.txt", package = 'Giotto')
loc_path = system.file("extdata", "seqfish_field_locs.txt", package = 'Giotto')
VC_small <- createGiottoObject(raw_exprs = expr_path, spatial_locs = loc_path)
# 2. normalize giotto
VC_small <- normalizeGiotto(gobject = VC_small, scalefactor = 6000)
VC_small <- addStatistics(gobject = VC_small)
# 3. dimension reduction</pre>
```

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```
VC_small <- calculateHVG(gobject = VC_small)
VC_small <- runPCA(gobject = VC_small)
VC_small <- runUMAP(VC_small, dimensions_to_use = 1:5, n_threads = 2)
plotUMAP(gobject = VC_small)</pre>
```

screePlot

screePlot

### **Description**

identify significant prinicipal components (PCs) using an screeplot (a.k.a. elbowplot)

## Usage

```
screePlot(
 gobject,
 name = "pca",
 expression_values = c("normalized", "scaled", "custom"),
 reduction = c("cells", "genes"),
 method = c("irlba", "factominer"),
 rev = FALSE,
 genes_to_use = NULL,
 center = F,
  scale_unit = F,
 ncp = 100,
 ylim = c(0, 20),
  verbose = T,
  show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "screePlot"
)
```

```
gobject
                  giotto object
                  name of PCA object if available
name
expression_values
                  expression values to use
reduction
                  cells or genes
method
                  which implementation to use
rev
                  do a reverse PCA
genes_to_use
                  subset of genes to use for PCA
center
                  center data before PCA
scale_unit
                  scale features before PCA
                  number of principal components to calculate
ncp
ylim
                  y-axis limits on scree plot
```

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```
verobsity
verbose
                  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
```

### **Details**

Screeplot works by plotting the explained variance of each individual PC in a barplot allowing you to identify which PC provides a significant contribution (a.k.a 'elbow method').

Screeplot will use an available pca object, based on the parameter 'name', or it will create it if it's not available (see runPCA)

#### Value

ggplot object for scree method

### **Examples**

```
screePlot(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

```
Output from detectSpatialPatterns
dimensions
                  dimensions to identify correlated genes for.
                  Top positively correlated genes.
top_pos_genes
                  Top negatively correlated genes.
top_neg_genes
min_pos_cor
                  Minimum positive correlation score to include a gene.
                  Minimum negative correlation score to include a gene.
min_neg_cor
```

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### **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

```
select_spatialNetwork
```

# Description

function to select a spatial network

```
select_spatialNetwork(gobject, name = NULL, return_network_Obj = FALSE)
```

## Description

sets the python path and/or install miniconda and the python modules

## Usage

```
set_giotto_python_path(
   python_path = NULL,
   packages_to_install = c("pandas", "networkx", "python-igraph", "leidenalg",
        "python-louvain", "python.app", "scikit-learn")
)
```

show,giotto-method

show method for giotto class

# Description

show method for giotto class

# Usage

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

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

### **Description**

Creates dendrogram for selected clusters.

```
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,
```

showClusterHeatmap 269

```
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
cor
                  correlation score to calculate distance
                  distance method to use for hierarchical clustering
distance
h
                  height of horizontal lines to plot
                  color of horizontal lines
h_color
rotate
                  rotate dendrogram 90 degrees
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters, see showSaveParameters
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

additional parameters for ggdendrogram()

# Details

Expression correlation dendrogram for selected clusters.

#### Value

ggplot

### **Examples**

showClusterDendrogram(gobject)

showClusterHeatmap showClusterHeatmap

# Description

Creates heatmap based on identified clusters

270 showClusterHeatmap

#### 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]
save_param
                  list of saving parameters, see showSaveParameters
default_save_name
```

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

additional parameters for the Heatmap function from ComplexHeatmap

### **Details**

Correlation heatmap of selected clusters.

#### Value

ggplot

# **Examples**

```
showClusterHeatmap(gobject)
```

 $show {\tt GiottoImageNames} \quad \textit{show GiottoImageNames}$ 

## Description

Prints the available giotto images that are attached to the Giotto object

### Usage

```
showGiottoImageNames(gobject, verbose = TRUE)
```

### **Arguments**

gobject a giotto object

verbose verbosity of function

### Value

a vector of giotto image names attached to the giotto object

### **Examples**

showGiottoImageNames(gobject)

 ${\tt showGiottoInstructions}$ 

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()

272 showNetworks

showGrids

showGrids

## Description

Prints the available spatial grids that are attached to the Giotto object

## Usage

```
showGrids(gobject, verbose = TRUE)
```

## Arguments

gobject

a giotto object

verbose

verbosity of function#'

## Value

vector

# **Examples**

showGrids()

showNetworks

showNetworks

# Description

Prints the available spatial networks that are attached to the Giotto object

### Usage

```
showNetworks(gobject, verbose = TRUE)
```

## **Arguments**

gobject

a giotto object

verbose

verbosity of function#'

#### Value

vector

# **Examples**

showNetworks()

showPattern 273

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)

274 showPattern2D

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**

```
showPattern2D(gobject)
```

showPattern3D 275

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"
```

```
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
```

276 showPatternGenes

#### 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"
)
```

```
gobject
                  giotto object
                  Output from detectSpatialPatterns
spatPatObj
dimension
                  dimension to plot genes for.
                  Top positively correlated genes.
top_pos_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
```

showProcessingSteps 277

```
return_plot return ggplot object
```

save\_plot directly save the plot [boolean]

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

default\_save\_name

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

### Value

ggplot

## **Examples**

showPatternGenes(gobject)

showProcessingSteps showProcessingSteps

# Description

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

#### Usage

showProcessingSteps(gobject)

## **Arguments**

gobject giotto object

#### Value

list of processing steps and names

### **Examples**

showProcessingSteps(gobject)

278 showSpatialCorGenes

```
showSaveParameters showSaveParameters
```

## Description

Description of Giotto saving options, links to all\_plots\_save\_function

#### Usage

```
showSaveParameters()
```

#### Value

Instruction on how to use the automatic plot saving options within Giotto

#### **Examples**

```
showSaveParameters()
```

showSpatialCorGenes show

show Spatial Cor Genes

## Description

Shows and filters spatially correlated genes

### Usage

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

signPCA 279

```
min_cor_diff filter on minimum correlation difference (spatial vs expression)
min_rank_diff filter on minimum correlation rank difference (spatial vs expression)
show_top_genes show top genes per gene
```

#### Value

data.table with filtered information

## **Examples**

```
showSpatialCorGenes(gobject)
```

signPCA

signPCA

### **Description**

identify significant prinicipal components (PCs)

# Usage

```
signPCA(
 gobject,
 name = "pca",
 method = c("screeplot", "jackstraw"),
 expression_values = c("normalized", "scaled", "custom"),
 reduction = c("cells", "genes"),
 pca_method = c("irlba", "factominer"),
 rev = FALSE,
 genes_to_use = NULL,
 center = T,
  scale_unit = T,
 ncp = 50,
 scree_ylim = c(0, 10),
  jack_iter = 10,
  jack_threshold = 0.01,
  jack_ylim = c(0, 1),
 verbose = TRUE,
  show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "signPCA"
)
```

```
gobject giotto object

name name of PCA object if available

method method to use to identify significant PCs
```

280 signPCA

expression\_values

expression values to use

reduction cells or genes

pca\_method which implementation to use

rev do a reverse PCA

genes\_to\_use subset of genes to use for PCA

center center data before PCA

scale\_unit scale features before PCA

ncp number of principal components to calculate

scree\_ylim y-axis limits on scree plot

jack\_iter number of interations for jackstraw

jack\_threshold p-value threshold to call a PC significant

jack\_ylim y-axis limits on jackstraw plot

verbose verbosity 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**

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 provides a significant contribution (a.k.a. 'elbow method').
- 2. The Jackstraw method uses the permutationPA function. By systematically permuting genes it identifies robust, and thus significant, PCs.

#### Value

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

### **Examples**

signPCA(gobject)

silhouetteRank 281

silhouetteRank	silhouetteRank
SITHOUGULENAIN	simoueneran

### **Description**

Previously: calculate\_spatial\_genes\_python. This method computes a silhouette score per gene based on the spatial distribution of two partitions of cells (expressed L1, and non-expressed L0). Here, rather than L2 Euclidean norm, it uses a rank-transformed, exponentially weighted function to represent the local physical distance between two cells.

# Usage

```
silhouetteRank(
  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
```

### Value

data.table with spatial scores

## **Examples**

```
silhouetteRank(gobject)
```

282 spatCellCellcom

```
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

spatCellCellcom

#### **Description**

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

### Usage

spatCellPlot 283

```
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
adjust_method
                  which method to adjust p-values
                  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.

#### Value

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

#### **Examples**

```
spatCellCellcom(gobject)
```

spatCellPlot spatCellPlot

### **Description**

Visualize cells according to spatial coordinates

```
spatCellPlot(
  gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  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,
```

284 spatCellPlot

```
point_shape = c("border", "no_border", "voronoi"),
 point_size = 3,
 point_alpha = 1,
 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 = NULL,
 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",
 vor_border_color = "white",
  vor_max_radius = 200,
 axis_text = 8,
 axis_title = 8,
 cow_n_col = 2,
 cow_rel_h = 1,
 cow_rel_w = 1,
 cow_align = "h",
 show_plot = NA,
 return_plot = NA,
 save_plot = NA,
 save_param = list(),
 default_save_name = "spatCellPlot"
)
```

```
gobject giotto object
show_image show a tissue background image
gimage a giotto image
image_name name of a giotto image
sdimx x-axis dimension name (default = 'sdimx')
sdimy y-axis dimension name (default = 'sdimy')
```

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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 shape of points (border, no\_border or voronoi) point\_size size of point (cell) point\_alpha transparancy of spatial points point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network alpha of spatial network network\_alpha 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

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```
show_legend
                  show legend
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
                  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
vor_alpha
                  transparancy of voronoi 'cells'
```

#### **Details**

Description of parameters.

# Value

ggplot

# **Examples**

spatCellPlot(gobject)

spatCellPlot2D spatCellPlot2D

### **Description**

Visualize cells according to spatial coordinates

```
spatCellPlot2D(
  gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
```

spatCellPlot2D 287

```
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", "voronoi"),
 point_size = 3,
 point_alpha = 1,
 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 = NULL,
 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",
 vor_border_color = "white",
 vor_max_radius = 200,
 vor_alpha = 1,
 axis_text = 8,
 axis_title = 8,
 cow_n_col = 2,
 cow_rel_h = 1,
 cow_rel_w = 1,
  cow_align = "h",
  show_plot = NA,
 return_plot = NA,
  save_plot = NA,
  save_param = list(),
 default_save_name = "spatCellPlot2D"
)
```

#### **Arguments**

gobject giotto object

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show\_image show a tissue background image a giotto image gimage name of a giotto image image\_name sdimx x-axis dimension name (default = 'sdimx') sdimy y-axis dimension name (default = 'sdimy') spat\_enr\_names names of spatial enrichment results to include cell\_annotation\_values numeric cell annotation columns cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits select\_cell\_groups select subset of cells/clusters based on cell\_color parameter select\_cells select subset of cells based on cell IDs point\_shape shape of points (border, no\_border or voronoi) point\_size size of point (cell) point\_alpha transparancy of spatial points point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network alpha of spatial network network\_alpha 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
show_legend
                  show legend
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
vor_alpha
                  transparancy of voronoi 'cells'
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]
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)

spatDimCellPlot
 spatDimCellPlot

# Description

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

```
spatDimCellPlot(
  gobject,
  show_image = F,
 gimage = NULL,
  image_name = "image",
 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_alpha = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border", "voronoi"),
  spat_point_size = 1,
  spat_point_alpha = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
  dim_show_center_label = T,
  dim_center_point_size = 4,
  dim_center_point_border_col = "black",
 dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_center_point_border_col = "black",
  spat_center_point_border_stroke = 0.1,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 nn_network_name = "sNN.pca",
  dim_edge_alpha = 0.5,
  spat_show_network = F,
  spatial_network_name = "Delaunay_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",
  vor_border_color = "white",
  vor_max_radius = 200,
  vor_alpha = 1,
  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
                  show a tissue background image
show_image
gimage
                  a giotto image
                  name of a giotto image
image_name
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
                  dimension to use on x-axis
dim1_to_use
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
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 spatial points with border or not (border or no\_border) dim\_point\_size size of points in dim. reduction space dim\_point\_alpha transparancy of dim. reduction points 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 shape of points (border, no\_border or voronoi) spat\_point\_size size of spatial points spat\_point\_alpha transparancy 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 coord\_fix\_ratio ratio for coordinates cowplot param: how many columns cow\_n\_col cowplot param: relative height cow\_rel\_h cowplot param: relative width cow\_rel\_w cow\_align cowplot param: how to align show\_legend show legend legend\_text size of legend text 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 vor\_border\_color border colorr for voronoi plot vor\_max\_radius maximum radius for voronoi 'cells'

```
vor_alpha
                  transparancy of voronoi 'cells'
                  size of axis text
axis_text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

#### **Details**

Description of parameters.

#### Value

ggplot

# **Examples**

```
spatDimCellPlot(gobject)
```

spatDimCellPlot2D

## **Description**

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

```
spatDimCellPlot2D(
 gobject,
  show_image = F,
 gimage = NULL,
  image_name = "image",
 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_alpha = 1,
dim_point_border_col = "black",
dim_point_border_stroke = 0.1,
spat_point_shape = c("border", "no_border", "voronoi"),
spat_point_size = 1,
spat_point_alpha = 1,
spat_point_border_col = "black",
spat_point_border_stroke = 0.1,
dim_show_cluster_center = F,
dim_show_center_label = T,
dim_center_point_size = 4,
dim_center_point_border_col = "black",
dim_center_point_border_stroke = 0.1,
dim_label_size = 4,
dim_label_fontface = "bold",
spat_show_cluster_center = F,
spat_show_center_label = F,
spat_center_point_size = 4,
spat_center_point_border_col = "black",
spat_center_point_border_stroke = 0.1,
spat_label_size = 4,
spat_label_fontface = "bold",
show_NN_network = F,
nn_network_to_use = "sNN",
nn_network_name = "sNN.pca",
dim_edge_alpha = 0.5,
spat\_show\_network = F,
spatial_network_name = "Delaunay_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",
vor_border_color = "white",
vor_max_radius = 200,
vor_alpha = 1,
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
                      show a tissue background image
    show_image
    gimage
                      a giotto image
                      name of a giotto image
    image_name
    plot_alignment direction to align plot
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                      numeric cell annotation columns
    {\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
    sdimx
                      = spatial dimension to use on x-axis
                      = spatial dimension to use on y-axis
    sdimy
    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_alpha
                      transparancy of dim. reduction points
    dim_point_border_col
                      border color of points in dim. reduction space
```

border stroke of points in dim. reduction space

dim\_point\_border\_stroke

spat\_point\_shape shape of points (border, no\_border or voronoi) spat\_point\_size size of spatial points spat\_point\_alpha transparancy 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 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
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
vor_alpha
                  transparancy of voronoi 'cells'
axis_text
                  size of axis text
                  size of axis title
axis_title
coord_fix_ratio
                  ratio for coordinates
cow_n_col
                  cowplot param: how many columns
                  cowplot param: relative height
cow_rel_h
                  cowplot param: relative width
cow_rel_w
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

### **Details**

Description of parameters.

## Value

ggplot

### **Examples**

spatDimCellPlot2D(gobject)

spatDimGenePlot 299

spatDimGenePlot

spatDimGenePlot

### **Description**

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

```
spatDimGenePlot(
 gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  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_alpha = 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,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spatial_network_name = "Delaunay_network",
  spatial_grid_name = "spatial_grid",
  spat_point_shape = c("border", "no_border", "voronoi"),
  spat_point_size = 1,
  spat_point_alpha = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
  show_legend = T,
  legend_text = 8,
 dim_background_color = "white",
  spat_background_color = "white",
  vor_border_color = "white",
  vor_max_radius = 200,
```

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```
vor_alpha = 1,
      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
    show_image
                     show a tissue background image
                     a giotto image
    gimage
    image_name
                     name of a giotto image
    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
                     dimension to use on y-axis
   dim2_to_use
    dim_point_shape
                     dim reduction points with border or not (border or no_border)
    dim_point_size dim reduction plot: point size
    dim_point_alpha
                     transparancy of dim. reduction points
    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 expression with ggplot alpha parameter

scale\_alpha\_with\_expression

sdimx spatial x-axis dimension name (default = 'sdimx') sdimy spatial y-axis dimension name (default = 'sdimy') 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\_alpha transparancy of spatial points spat\_point\_border\_col color of border around points spat\_point\_border\_stroke stroke size of border around points cell\_color\_gradient vector with 3 colors for numeric data gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits 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 vor\_border\_color border colorr for voronoi plot vor\_max\_radius maximum radius for voronoi 'cells' vor\_alpha transparancy of voronoi 'cells' 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 cow\_rel\_w cowplot param: relative width cowplot param: how to align cow\_align show plots 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 spatDimGenePlot2D

#### **Details**

Description of parameters.

#### Value

ggplot

## See Also

spatDimGenePlot3D

### **Examples**

```
spatDimGenePlot(gobject)
```

spatDimGenePlot2D

spatDimGenePlot2D

## **Description**

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

```
spatDimGenePlot2D(
 gobject,
  show_image = F,
 gimage = NULL,
  image_name = "image",
 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_alpha = 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,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spatial_network_name = "Delaunay_network",
```

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```
spatial_grid_name = "spatial_grid",
      spat_point_shape = c("border", "no_border", "voronoi"),
      spat_point_size = 1,
      spat_point_alpha = 1,
      spat_point_border_col = "black",
      spat_point_border_stroke = 0.1,
      cell_color_gradient = c("blue", "white", "red"),
      gradient_midpoint = NULL,
      gradient_limits = NULL,
      cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_legend = T,
      legend_text = 8,
      dim_background_color = "white",
      spat_background_color = "white",
      vor_border_color = "white",
      vor_max_radius = 200,
      vor_alpha = 1,
      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
                    show a tissue background image
    show_image
    gimage
                    a giotto image
                    name of a giotto image
    image_name
    expression_values
                    gene expression values to use
   plot_alignment direction to align plot
    genes
                    genes to show
    dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
   dim1_to_use
                    dimension to use on x-axis
    dim2_to_use
                    dimension to use on y-axis
    dim_point_shape
                    dim reduction points with border or not (border or no_border)
    dim_point_size dim reduction plot: point size
    dim_point_alpha
```

transparancy of dim. reduction points

```
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
sdimx
                 spatial x-axis dimension name (default = 'sdimx')
sdimy
                 spatial y-axis dimension name (default = 'sdimy')
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_alpha
                 transparancy of spatial points
spat_point_border_col
                 color of border around points
spat_point_border_stroke
                 stroke size of border around points
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
                 cowplot param: how many columns
cow_n_col
                 cowplot param: relative height
cow_rel_h
                 cowplot param: relative width
cow_rel_w
cow_align
                 cowplot param: how to align
show_legend
                 show legend
legend_text
                 size of legend text
dim_background_color
                  color of plot background for dimension plot
spat_background_color
                 color of plot background for spatial plot
vor_border_color
                 border colorr for voronoi plot
```

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```
vor_max_radius maximum radius for voronoi 'cells'
vor_alpha
                  transparancy of voronoi 'cells'
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plots
                  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
```

# **Details**

Description of parameters.

#### Value

ggplot

#### See Also

```
spatDimGenePlot3D
```

### **Examples**

```
spatDimGenePlot2D(gobject)
```

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,
```

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```
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 = "Delaunay_network",
 network_color = "lightgray",
  spatial_network_alpha = 0.5,
  show_spatial_grid = F,
  spatial_grid_name = "spatial_grid",
  spatial_grid_color = NULL,
  spatial_grid_alpha = 0.5,
  spatial_point_size = 3,
  legend_text_size = 12,
  axis_scale = c("cube", "real", "custom"),
 custom_ratio = NULL,
 x_{ticks} = NULL,
 y_ticks = NULL,
 z_ticks = NULL,
  show_plot = NA,
 return_plot = NA,
  save_plot = NA,
 save_param = list(),
 default_save_name = "spatDimGenePlot3D"
)
```

# **Arguments**

```
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 to show
genes
```

```
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
                 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
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,
  show_image = F,
 gimage = NULL,
  image_name = "image",
 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_alpha = 1,
 dim_point_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border", "voronoi"),
  spat_point_size = 1,
  spat_point_alpha = 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 = NULL,
  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",
  vor_border_color = "white",
  vor_max_radius = 200,
  vor_alpha = 1,
  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
show_image
                  show a tissue background image
                  a giotto image
gimage
image_name
                  name of a giotto image
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
                  = spatial dimension to use on x-axis
sdimx
                  = 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\_alpha transparancy of point 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 shape of points (border, no\_border or voronoi) spat\_point\_size size of spatial points spat\_point\_alpha transparancy 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 vor\_border\_color border colorr for voronoi plot vor\_max\_radius maximum radius for voronoi 'cells' vor\_alpha transparancy of voronoi 'cells' axis\_text size of axis text

axis\_title

size of axis title

```
show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

### **Details**

Description of parameters.

### Value

ggplot

### See Also

```
spatDimPlot2D and spatDimPlot3D for 3D visualization.
```

### **Examples**

```
spatDimPlot(gobject)
```

spatDimPlot2D spatDimPlot2D

## **Description**

Visualize cells according to spatial AND dimension reduction coordinates 2D

```
spatDimPlot2D(
 gobject,
  show_image = F,
 gimage = NULL,
  image_name = "image",
 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_alpha = 1,
dim_point_border_col = "black",
dim_point_border_stroke = 0.1,
spat_point_shape = c("border", "no_border", "voronoi"),
spat_point_size = 1,
spat_point_alpha = 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 = NULL,
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",
vor_border_color = "white",
vor_max_radius = 200,
vor_alpha = 1,
axis_text = 8,
axis_title = 8,
show_plot = NA,
```

```
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spatDimPlot2D"
)
```

## **Arguments**

```
gobject
                  giotto object
                  show a tissue background image
show_image
                  a giotto image
gimage
image_name
                  name of a giotto image
plot_alignment direction to align plot
dim_reduction_to_use
                  dimension reduction to use
dim_reduction_name
                  dimension reduction name
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
sdimx
                  = spatial dimension to use on x-axis
sdimy
                  = spatial dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
                  color for cells (see details)
cell_color
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
                  select subset of cells based on cell IDs
select_cells
dim_point_shape
                  point with border or not (border or no_border)
dim_point_size size of points in dim. reduction space
dim_point_alpha
                  transparancy of point 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
                  shape of points (border, no_border or voronoi)
```

spat\_point\_size size of spatial points spat\_point\_alpha transparancy 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

color of spatial grid

spat\_grid\_color

```
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
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
                  transparancy of voronoi 'cells'
vor_alpha
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

## **Details**

Description of parameters.

### Value

ggplot

## See Also

spatDimPlot3D

### **Examples**

spatDimPlot2D(gobject)

spatDimPlot3D

spatDimPlot3D

## **Description**

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

```
spatDimPlot3D(
 gobject,
 plot_alignment = c("horizontal", "vertical"),
  dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2\_to\_use = 2,
 dim3\_to\_use = 3,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  show_NN_network = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
  show_cluster_center = F,
  show_center_label = T,
 center_point_size = 4,
 label_size = 16,
  select_cell_groups = NULL,
  select_cells = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1.5,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
 dim_point_size = 3,
 nn_network_alpha = 0.5,
  show_spatial_network = F,
  spatial_network_name = "Delaunay_network",
 network_color = "lightgray",
  spatial_network_alpha = 0.5,
  show_spatial_grid = F,
  spatial_grid_name = "spatial_grid",
  spatial_grid_color = NULL,
  spatial_grid_alpha = 0.5,
  spatial_point_size = 3,
  axis_scale = c("cube", "real", "custom"),
  custom_ratio = NULL,
 x_ticks = NULL,
 y_{ticks} = NULL,
  z_ticks = NULL,
```

```
legend_text_size = 12,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimPlot3D"
Arguments
    gobject
                      giotto object
    plot_alignment direction to align plot
    dim_reduction_to_use
                      dimension reduction to use
    dim_reduction_name
                      dimension reduction name
    dim1_to_use
                      dimension to use on x-axis
    dim2_to_use
                      dimension to use on y-axis
                      dimension to use on z-axis
    dim3_to_use
    sdimx
                      = spatial dimension to use on x-axis
    sdimy
                      = spatial dimension to use on y-axis
                      = spatial dimension to use on z-axis
    sdimz
    show_NN_network
                      show underlying NN network
    nn_network_to_use
                      type of NN network to use (kNN vs sNN)
                      name of NN network to use, if show_NN_network = TRUE
    network_name
    show_cluster_center
                      show the center of each cluster
    show_center_label
                      provide a label for each cluster
    center_point_size
                      size of the center point
    label_size
                      size of the center label
    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
    cell_color
                      color for cells (see details)
    color_as_factor
                      convert color column to factor
    cell_color_code
                      named vector with colors
```

```
dim_point_size size of points in dim. reduction space
nn_network_alpha
                  column to use for alpha of the edges
show_spatial_network
                  show spatial network
spatial_network_name
                  name of spatial network to use
spatial_network_alpha
                  alpha of spatial network
show_spatial_grid
                  show spatial grid
spatial_grid_name
                  name of spatial grid to use
spatial_grid_color
                  color of spatial grid
spatial_point_size
                  size of spatial points
                  show plot
show_plot
return_plot
                  return ggplot object
                  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
dim_point_border_col
                  border color of points in dim. reduction space
dim_point_border_stroke
                  border stroke of points in dim. reduction space
spatial_network_color
                  color of spatial network
spatial_other_point_size
                  size of not selected spatial points
{\tt spatial\_other\_cells\_alpha}
                  alpha of not selected spatial points
dim_other_point_size
                  size of not selected dim. reduction points
show_legend
                  show legend
```

### **Details**

Description of parameters.

### Value

plotly

### **Examples**

spatDimPlot3D(gobject)

320 spatGenePlot

spatGenePlot

spatGenePlot

## **Description**

Visualize cells and gene expression according to spatial coordinates

```
spatGenePlot(
 gobject,
  show_image = F,
 gimage = NULL,
  image_name = "image",
  sdimx = "sdimx",
  sdimy = "sdimy",
  expression_values = c("normalized", "scaled", "custom"),
 genes,
  cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 show_network = F,
 network_color = NULL,
  spatial_network_name = NULL,
 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", "voronoi"),
 point_size = 1,
 point_alpha = 1,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_legend = T,
 legend_text = 8,
 background_color = "white",
 vor_border_color = "white",
  vor_max_radius = 200,
  vor_alpha = 1,
 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(),
```

spatGenePlot 321

```
default_save_name = "spatGenePlot"
)
```

### **Arguments**

gobject giotto object

show\_image show a tissue background image

gimage a giotto image

image\_name name of a giotto image

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

expression\_values

gene expression values to use

genes genes to show

cell\_color\_gradient

vector with 3 colors for numeric data

gradient\_midpoint

midpoint for color gradient

gradient\_limits

vector with lower and upper limits

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\_shape shape of points (border, no\_border or voronoi)

point\_size size of point (cell)
point\_alpha transparancy of points

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

vor\_border\_color

border colorr for voronoi plot

vor\_max\_radius maximum radius for voronoi 'cells'

322 spatGenePlot2D

```
vor_alpha
                  transparancy of voronoi 'cells'
                  size of axis text
axis_text
                  size of axis title
axis_title
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
                  cowplot param: relative width
cow_rel_w
                  cowplot param: how to align
cow_align
show_plot
                  show plots
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
                  parameters for cowplot::save_plot()
. . .
```

### **Details**

Description of parameters.

### Value

ggplot

## See Also

```
spatGenePlot3D and spatGenePlot2D
```

### **Examples**

```
spatGenePlot(gobject)
```

spatGenePlot2D spatGenePlot2D

## **Description**

Visualize cells and gene expression according to spatial coordinates

```
spatGenePlot2D(
  gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  sdimx = "sdimx",
  sdimy = "sdimy",
  expression_values = c("normalized", "scaled", "custom"),
```

spatGenePlot2D 323

```
genes,
cell_color_gradient = c("blue", "white", "red"),
gradient_midpoint = NULL,
gradient_limits = NULL,
show_network = F,
network_color = NULL,
spatial_network_name = NULL,
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", "voronoi"),
point_size = 1,
point_alpha = 1,
point_border_col = "black",
point_border_stroke = 0.1,
show_legend = T,
legend_text = 8,
background_color = "white",
vor_border_color = "white",
vor_alpha = 1,
vor_max_radius = 200,
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
                   show a tissue background image
show_image
gimage
                   a giotto image
image_name
                   name of a giotto image
\operatorname{sdim} x
                   x-axis dimension name (default = 'sdimx')
                   y-axis dimension name (default = 'sdimy')
sdimy
expression_values
                   gene expression values to use
genes
                   genes to show
cell_color_gradient
                   vector with 3 colors for numeric data
```

324 spatGenePlot2D

gradient\_midpoint midpoint for color gradient gradient\_limits vector with lower and upper limits show\_network show underlying spatial network color of spatial network network\_color spatial\_network\_name name of spatial network to use show\_grid show spatial grid grid\_color color of spatial grid spatial\_grid\_name name of spatial grid to use midpoint expression midpoint scale\_alpha\_with\_expression scale expression with ggplot alpha parameter shape of points (border, no\_border or voronoi) point\_shape size of point (cell) point\_size transparancy of points point\_alpha 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 vor\_border\_color border colorr for voronoi plot transparancy of voronoi 'cells' vor\_alpha vor\_max\_radius maximum radius for voronoi 'cells' 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 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() . . .

spatGenePlot3D 325

#### **Details**

Description of parameters.

#### Value

ggplot

## See Also

spatGenePlot3D

## **Examples**

```
spatGenePlot2D(gobject)
```

spatGenePlot3D

spatGenePlot3D

## **Description**

Visualize cells and gene expression according to spatial coordinates

```
spatGenePlot3D(
 gobject,
 expression_values = c("normalized", "scaled", "custom"),
 genes,
 show_network = F,
 network_color = NULL,
  spatial_network_name = "Delaunay_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,
```

326 spatGenePlot3D

```
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spatGenePlot3D"
)
```

# Arguments

```
giotto object
gobject
expression_values
                 gene expression values to use
                 genes to show
genes
                 show underlying spatial network
show_network
network_color
                 color of spatial network
spatial_network_name
                 name of spatial network to use
show_grid
                 show spatial grid
genes_high_color
                 color represents high gene expression
genes_mid_color
                 color represents middle gene expression
genes_low_color
                 color represents low gene expression
spatial_grid_name
                 name of spatial grid to use
point_size
                 size of point (cell)
show_legend
                 show legend
show_plot
                 show plots
return_plot
                 return ggplot object
save_plot
                 directly save the plot [boolean]
                 list of saving parameters from all_plots_save_function
save_param
default_save_name
                 default save name for saving, don't change, change save_name in save_param
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

```
spatGenePlot3D(gobject)
```

spatialAEH 327

spatialAEH

spatialAEH

# Description

Compute spatial variable genes with spatialDE method

# 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**

```
gobject Giotto object

SpatialDE_results
results of SpatialDE function

name_pattern name for the computed spatial patterns
expression_values
gene expression values to use

pattern_num number of spatial patterns to look for

lengthscale

python_path specify specific path to python if required

return_gobject show plot
```

# **Details**

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

#### Value

An updated giotto object

```
spatialAEH(gobject)
```

328 spatialDE

spatialDE	spatialDE
Spatiaide	spanabl

#### **Description**

Compute spatial variable genes with spatialDE method

#### 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

```
gobject
                  Giotto object
expression_values
                  gene expression values to use
                  size of plot
size
color
                  low/medium/high color scheme for plot
                  alpha value for significance
sig_alpha
                  alpha value for unsignificance
unsig_alpha
python_path
                  specify specific path to python if required
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**

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

#### Value

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

Spatial\_AEH 329

## **Examples**

```
spatialDE(gobject)
```

Spatial\_AEH

Spatial\_AEH

## **Description**

calculate automatic expression histology with spatialDE method

# Usage

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

## **Arguments**

gobject Giotto object
results output from spatial\_DE
pattern\_num the number of gene expression patterns
show\_AEH show AEH plot
python\_path specify specific path to python if required

Details

Description.

#### Value

a list or a dataframe of SVs

```
Spatial_AEH(gobject)
```

330 spatNetwDistributions

Spatial\_DE

Spatial\_DE

# Description

calculate spatial varible genes with spatialDE method

# Usage

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

# Arguments

gobject Giotto object show\_plot show FSV plot

python\_path specify specific path to python if required

# Details

Description.

## Value

a list or a dataframe of SVs

# **Examples**

```
Spatial_DE(gobject)
```

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

# Description

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

spatNetwDistributions 331

#### **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
                  number of binds to use for the histogram
hist_bins
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)
```

334 spatPlot

spatPlot spatPlot

#### **Description**

Visualize cells according to spatial coordinates

```
spatPlot(
 gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
 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", "voronoi"),
 point_size = 3,
 point_alpha = 1,
 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 = NULL,
 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,
```

spatPlot 335

```
title = NULL,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
  background_color = "white",
  vor_border_color = "white",
  vor_max_radius = 200,
  vor_alpha = 1,
  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
                  show a tissue background image
show_image
                  a giotto image
gimage
                  name of a giotto image
image_name
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_cells
                  select subset of cells based on cell IDs
                  shape of points (border, no_border or voronoi)
point_shape
point_size
                  size of point (cell)
point_alpha
                  transparancy of point
```

336 spatPlot

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 color of spatial network network\_color network\_alpha alpha of spatial network show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use grid\_color color of spatial grid show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size point size of not selected cells other\_cells\_alpha alpha of not selected cells coord\_fix\_ratio fix ratio between x and y-axis 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 vor\_border\_color border colorr for voronoi plot vor\_max\_radius maximum radius for voronoi 'cells'

transparancy of voronoi 'cells'

cowplot param: how many columns

size of axis text

size of axis title

vor\_alpha

axis\_text

axis\_title

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_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
                  create multiple plots based on cell annotation column
groub_by
```

## **Details**

Description of parameters.

#### Value

ggplot

#### See Also

spatPlot3D

#### **Examples**

spatPlot(gobject)

spatPlot2D

spatPlot2D

## **Description**

Visualize cells according to spatial coordinates

```
spatPlot2D(
  gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  group_by = NULL,
  group_by_subset = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  spat_enr_names = NULL,
  cell_color = NULL,
  color_as_factor = T,
  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", "voronoi"),
 point_size = 3,
 point_alpha = 1,
 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 = NULL,
 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",
 vor_border_color = "white",
 vor_max_radius = 200,
 vor_alpha = 1,
 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

show\_image show a tissue background image a giotto image gimage name of a giotto image image\_name 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 subset of cells based on cell IDs select\_cells shape of points (border, no\_border or voronoi) point\_shape point\_size size of point (cell) point\_alpha transparancy of point point\_border\_col color of border around points point\_border\_stroke stroke size of border around points show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label\_size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network network\_alpha alpha of spatial network show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use

```
grid_color
                  color of spatial grid
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
                  title of plot
title
show_legend
                  show legend
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
vor_alpha
                  transparancy of voronoi 'cells'
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
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
                  return ggplot object
return_plot
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
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.

#### Value

ggplot

#### See Also

spatPlot3D

spatPlot2D\_single 341

#### **Examples**

```
spatPlot2D(gobject)
```

## **Description**

Visualize cells according to spatial coordinates

```
spatPlot2D_single(
 gobject,
  show_image = F,
  gimage = NULL,
  image_name = "image",
  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", "voronoi"),
 point_size = 3,
 point_alpha = 1,
 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 = NULL,
 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,
```

342 spatPlot2D\_single

```
other_cells_alpha = 0.1,
      coord_fix_ratio = NULL,
      title = NULL,
      show_legend = T,
      legend_text = 8,
      legend_symbol_size = 1,
      background_color = "white",
      vor_border_color = "white",
      vor_max_radius = 200,
      vor_alpha = 1,
      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
                     show a tissue background image
    show_image
    gimage
                     a giotto image
    image_name
                     name of a giotto image
    sdimx
                     x-axis dimension name (default = 'sdimx')
    sdimy
                     y-axis dimension name (default = 'sdimy')
    spat_enr_names names of spatial enrichment results to include
                     color for cells (see details)
    cell_color
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
                     shape of points (border, no_border or voronoi)
   point_shape
                     size of point (cell)
    point_size
    point_alpha
                     transparancy of point
    point_border_col
                     color of border around points
   point_border_stroke
```

stroke size of border around points

show\_cluster\_center plot center of selected clusters show\_center\_label plot label of selected clusters center\_point\_size size of center points label size size of labels label\_fontface font of labels show\_network show underlying spatial network spatial\_network\_name name of spatial network to use network\_color color of spatial network network\_alpha alpha of spatial network show\_grid show spatial grid spatial\_grid\_name name of spatial grid to use grid\_color color of spatial grid show\_other\_cells display not selected cells other\_cell\_color color of not selected cells other\_point\_size point size of not selected cells other\_cells\_alpha alpha of not selected cells coord\_fix\_ratio fix ratio between x and y-axis title title of plot show\_legend show legend legend\_text size of legend text legend\_symbol\_size size of legend symbols background\_color color of plot background vor\_border\_color

border colorr for voronoi plot

vor\_max\_radius maximum radius for voronoi 'cells'

vor\_alpha transparancy of voronoi 'cells'

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

344 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 = "Delaunay_network",
 show_grid = F,
 grid_color = NULL,
 spatial_grid_name = "spatial_grid",
  title = "",
  show_legend = T,
 axis_scale = c("cube", "real", "custom"),
 custom_ratio = NULL,
 x_{ticks} = NULL,
 y_ticks = NULL,
```

spatPlot3D 345

```
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
```

346 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)
```

 ${\tt spat\_OR\_func}$ 

 $spat\_OR\_func$ 

# 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 = "Delaunay_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
                  first cell type
cell_type_1
cell_type_2
                  second cell type
gene_set_1
                  first specific gene set from gene pairs
                  second specific gene set from gene pairs
gene_set_2
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
                  which method to adjust p-values
adjust_method
adjust_target
                  adjust multiple hypotheses at the cell or gene level
verbose
                  verbose
```

348 standardise\_giotto

#### **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)
```

 ${\tt standardise\_giotto}$ 

standardise\_giotto

#### **Description**

```
standardises a matrix
```

```
standardise_giotto(x, center = TRUE, scale = TRUE)
```

stitchFieldCoordinates 349

#### **Arguments**

```
x matrix
center center data
scale scale data
```

#### Value

standardized matrix

```
stitchFieldCoordinates
```

stitchFieldCoordinates

## **Description**

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

#### Usage

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

# 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?
```

350 stitchTileCoordinates

#### **Details**

Stitching of fields:

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

#### Value

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

#### **Examples**

```
stitchFieldCoordinates(gobject)
```

```
stitchTileCoordinates
```

# Description

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

## Usage

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

## **Arguments**

location\_file location dataframe with X and Y coordinates

Xtilespan numerical value specifying the width of each tile

Ytilespan numerical value specifying the height of each tile

# **Details**

•••

```
\verb|stitchTileCoordinates(gobject)|\\
```

subClusterCells 351

subClusterCells

subClusterCells

#### **Description**

subcluster cells

#### Usage

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

## Arguments

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

352 subsetGiotto

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

pca\_param parameters for runPCA

nn\_param parameters for parameters for createNearestNetwork

 $k\_neighbors$  number of k for createNearestNetwork

resolution resolution gamma gamma omega omega

python\_path specify specific path to python if required

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

network\_name name of NN network to use

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

verbose verbose

## **Details**

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

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

## Value

giotto object with new subclusters appended to cell metadata

#### See Also

 ${\tt doLouvainCluster\_multinet}, {\tt doLouvainCluster\_community} \ and \ @see also \ {\tt doLeidenCluster\_community} \ and \ @see also \ {\tt doLeidenCluster\_community} \ and \ (a) \ {\tt doLouvainCluster\_community} \ and \ (b) \ {\tt doLouvainCluster\_community} \ and \ (c) \ {\tt doLouvainCluster\_com$ 

#### **Examples**

```
subClusterCells(gobject)
```

subsetGiotto

subsetGiotto

## **Description**

subsets Giotto object including previous analyses.

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

subsetGiottoLocs 353

## **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)
```

 $\verb"subsetGiottoLocs"$ 

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

354 trendSceek

#### Value

```
giotto object
```

# **Examples**

```
subsetGiottoLocs(gobject)
```

```
transform\_2d\_mesh\_to\_3d\_mesh\\ transform\_2d\_mesh\_to\_3d\_mesh
```

# Description

transform 2d mesh to 3d mesh by reversing PCA

# Usage

```
transform_2d_mesh_to_3d_mesh(
  mesh_line_obj_2d,
  pca_out,
  center_vec,
  mesh_grid_n
)
```

trendSceek

trendSceek

# Description

Compute spatial variable genes with trendsceek method

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

t\_giotto 355

## **Arguments**

```
gobject Giotto object
expression_values
gene expression values to use
subset_genes subset of genes to run trendsceek on
nrand An integer specifying the number of random resamplings of the mark distribution as to create the null-distribution.

ncores An integer specifying the number of cores to be used by BiocParallel
... Additional parameters to the trendsceek_test function
```

## **Details**

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

## Value

data.frame with trendsceek spatial genes results

## **Examples**

```
trendSceek(gobject)
```

```
t\_giotto t\_giotto
```

## **Description**

t\_giotto

## Usage

```
t\_giotto(mymatrix)
```

```
update {\tt GiottoImage} \qquad update {\tt GiottoImage}
```

## **Description**

Updates the boundaries of a giotto image attached to a giotto object

```
updateGiottoImage(
  gobject,
  image_name,
  xmax_adj = 0,
  xmin_adj = 0,
  ymax_adj = 0,
  ymin_adj = 0,
  return_gobject = TRUE
)
```

356 viewHMRFresults

## **Arguments**

```
gobject giotto object
image_name spatial locations

xmax_adj adjustment of the maximum x-value to align the image

xmin_adj adjustment of the minimum x-value to align the image

ymax_adj adjustment of the maximum y-value to align the image

ymin_adj adjustment of the minimum y-value to align the image

return_gobject return a giotto object
```

#### Value

a giotto object or an updated giotto image if return\_gobject = F

# **Examples**

```
updateGiottoImage(gobject)
```

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 ...

viewHMRFresults2D 357

## Value

spatial plots with HMRF domains

## See Also

```
visPlot
```

## **Examples**

```
viewHMRFresults(gobject)
```

viewHMRFresults2D

viewHMRFresults2D

# Description

View results from doHMRF.

# Usage

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

# **Arguments**

```
gobject giotto object
```

HMRF output from doHMRF k number of HMRF domains

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)
```

358 viewHMRFresults3D

viewHMRFresults3D

viewHMRFresults3D

# Description

View results from doHMRF.

# Usage

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

# **Arguments**

```
gobject giotto object
```

HMRF output from doHMRF k number of HMRF domains

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

... paramters to visPlot()

# **Details**

Description ...

# Value

spatial plots with HMRF domains

# See Also

```
spatPlot3D
```

```
viewHMRFresults3D(gobject)
```

violinPlot 359

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
```

360 visDimGenePlot

#### Value

ggplot

#### **Examples**

```
violinPlot(gobject)
```

visDimGenePlot

visDimGenePlot

## **Description**

Visualize cells and gene expression according to dimension reduction coordinates

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

visDimGenePlot 361

### **Arguments**

gobject giotto object

expression\_values

gene expression values to use

genes genes to show

dim\_reduction\_to\_use

dimension reduction to use

dim\_reduction\_name

dimension reduction name

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

show\_NN\_network

show underlying NN network

nn\_network\_to\_use

type of NN network to use (kNN vs sNN)

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

edge\_alpha column to use for alpha of the edges

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

midpoint size of point (cell)

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

show\_legend show legend show\_plots show plots

# Details

Description of parameters.

#### Value

ggplot

# Examples

visDimGenePlot(gobject)

### **Description**

Visualize cells and gene expression according to dimension reduction coordinates

### Usage

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

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

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

### **Details**

Description of parameters.

### Value

ggplot

### **Examples**

```
visDimGenePlot_2D_ggplot(gobject)
```

# Description

Visualize cells and gene expression according to dimension reduction coordinates

#### Usage

```
visDimGenePlot_3D_plotly(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  dim3_to_use = 3,
  show_NN_network = F,
  nn_network_to_use = "sNN",
  network_name = "sNN.pca",
  network_color = "lightgray",
  edge_alpha = NULL,
  point_size = 1,
  genes_high_color = NULL,
  genes_mid_color = "white",
  genes_low_color = "blue",
  show_legend = T,
  show_plots = F
)
```

# **Arguments**

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

### **Details**

Description of parameters.

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### Value

ggplot

#### **Examples**

```
visDimGenePlot_3D_plotly(gobject)
```

visDimPlot

visDimPlot

### **Description**

Visualize cells according to dimension reduction coordinates

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

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```
save_folder = NULL,
      save_name = NULL,
      save_format = NULL,
      show_saved_plot = F,
    )
Arguments
    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
    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

return\_plot

show\_plot

save\_plot
save\_dir

show legend

return ggplot object

directly save the plot [boolean]

directory to save the plot

show plot

visDimPlot\_2D\_ggplot 367

### **Details**

Description of parameters.

### Value

ggplot or plotly

### **Examples**

```
visDimPlot(gobject)
```

```
visDimPlot_2D_ggplot visDimPlot_2D_ggplot
```

### Description

Visualize cells according to dimension reduction coordinates

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

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

size of center points

visDimPlot\_2D\_plotly 369

#### **Details**

Description of parameters.

#### Value

ggplot

### **Examples**

```
visDimPlot_2D_ggplot(gobject)
```

```
visDimPlot_2D_plotly visDimPlot_2D_plotly
```

### **Description**

Visualize cells according to dimension reduction coordinates

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

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

### **Arguments**

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

### **Details**

Description of parameters.

### Value

plotly

# **Examples**

```
visDimPlot_2D_plotly(gobject)
```

```
visDimPlot_3D_plotly
```

# **Description**

Visualize cells according to dimension reduction coordinates

### Usage

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

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

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

### **Details**

Description of parameters.

### Value

plotly

### **Examples**

```
visDimPlot_3D_plotly(gobject)
```

visForceLayoutPlot visForceLayoutPlot

# Description

Visualize cells according to forced layout algorithm coordinates

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

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

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

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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

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
)
```

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

visGenePlot\_3D\_plotly

```
genes_low_color
```

color represents low gene expression

show\_network show underlying spatial network

network\_color color of spatial network

spatial\_network\_name

name of spatial network to use

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

spatial\_grid\_name

name of spatial grid to use

midpoint expression midpoint

scale\_alpha\_with\_expression

scale expression with ggplot alpha parameter

point\_size size of point (cell)

point\_border\_col

color of border around points

point\_border\_stroke

stroke size of border around points

show\_legend show legend

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

show\_plots show plots

### **Details**

Description of parameters.

### Value

ggplot

### **Examples**

visGenePlot\_2D\_ggplot(gobject)

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

# Description

Visualize cells and gene expression according to spatial coordinates

#### Usage

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

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

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### **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,
)
```

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

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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,
)
```

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

visPlot\_2D\_ggplot 383

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**

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

384 visPlot\_2D\_plotly

```
visPlot_2D_plotly
```

### **Description**

Visualize cells according to spatial coordinates

#### Usage

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

```
gobject giotto object

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

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

point_size size of point (cell)

cell_color color for cells (see details)

cell_color_code

named vector with colors

color_as_factor

convert color column to factor
```

visPlot\_3D\_plotly 385

```
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,
stow_plot = F
```

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

visSpatDimGenePlot 387

#### **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,
```

visSpatDimGenePlot

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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
```

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
   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
    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
```

alpha of spatial network show\_spatial\_grid

spatial\_network\_alpha

show spatial grid

network\_color color of spatial/nn network

spatial\_grid\_name

name of spatial grid to use

spatial\_grid\_color

color of spatial grid

spatial\_grid\_alpha

alpha of spatial grid

legend\_text\_size

text size of legend

show\_legend show legend show plot show\_plot

### **Details**

Description of parameters.

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#### Value

plotly

#### **Examples**

```
visSpatDimPlot_3D(gobject)
```

visSpatDimPlot

visSpatDimPlot

### **Description**

integration of visSpatDimPlot\_2D and visSpatDimPlot\_3D

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

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

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

### **Details**

Description of parameters.

#### Value

ggplot or plotly

### **Examples**

```
visSpatDimPlot(gobject)
```

visSpatDimPlot\_2D
visSpatDimPlot\_2D

### **Description**

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

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

visSpatDimPlot\_2D

 $show_NN_network = F,$ 

cell\_color

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```
nn_network_to_use = "sNN",
     network_name = "sNN.pca",
      show\_cluster\_center = F,
      show_center_label = T,
      center_point_size = 4,
      label_size = 4,
      label_fontface = "bold",
      cell_color = NULL,
      color_as_factor = T,
      cell_color_code = NULL,
      select_cell_groups = NULL,
      select_cells = NULL,
      show_other_cells = T,
     other_cell_color = "lightgrey",
      dim_plot_mode = NULL,
     dim_point_size = 1,
     dim_point_border_col = "black",
     dim_point_border_stroke = 0.1,
     nn_network_alpha = 0.05,
      show_spatial_network = F,
      spatial_network_name = "spatial_network",
      spatial_network_color = NULL,
      show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_point_size = 1,
      spatial_point_border_col = "black",
      spatial_point_border_stroke = 0.1,
      show_legend = T,
      show_plot = F,
     plot_method = "ggplot"
Arguments
   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
   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)

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

#### **Details**

Description of parameters.

#### Value

ggplot

## **Examples**

```
visSpatDimPlot_2D(gobject)
```

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visSpatDimPlot\_3D vis

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

400 visSpatDimPlot\_3D

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

## **Details**

Description of parameters.

#### Value

plotly

writeHMRFresults 401

## **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

# Usage

```
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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