Package 'Giotto'

April 29, 2020

```
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      ggplot2 (>= 3.1.1),
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Imports Matrix,
      magick,
      uwot (>= 0.0.0.9010),
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      scales (>= 1.0.0),
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      igraph (>= 1.2.4.1),
      plotly,
      reticulate (>= 1.14),
      magrittr,
      limma,
      ggdendro,
      smfishHmrf,
      matrixStats (>= 0.55.0),
      devtools,
```

reshape2,

2 R topics documented:

```
ggraph,
     Rcpp,
     Rfast,
     Rtsne (>= 0.15),
      rlang (>= 0.4.3),
      fitdistrplus,
      deldir
Suggests knitr,
     rmarkdown,
     MAST,
      scran (>= 1.10.1),
     png,
     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:

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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.

Usage

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

Arguments

```
gobject giotto object

spatial_network

name of spatial network to use

cluster_column column of cell types

cell_interaction

cell-cell interaction to use

name name for the new metadata column

return_gobject return an updated giotto object
```

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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,
  by_column = FALSE,
  column_cell_ID = NULL
```

Arguments

gobject giotto object

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

by_column merge metadata based on cell_ID column in pDataDT (default = FALSE)

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

Details

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

Value

giotto object

Examples

```
addCellMetadata(gobject)
```

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addCellStatistics

addCellStatistics

Description

adds cells statistics to the giotto object

Usage

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

Arguments

Details

This function will add the following statistics to cell metadata:

- nr_genes: Denotes in how many genes are detected per cell
- perc_genes: Denotes what percentage of genes is detected per cell
- total_expr: Shows the total sum of gene expression per cell

Value

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

Examples

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

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addGeneMetadata

addGeneMetadata

Description

adds gene metadata to the giotto object

Usage

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

Arguments

```
gobject giotto object
```

new_metadata new metadata to use

by_column merge metadata based on gene_ID column in fDataDT column_cell_ID column name of new metadata to use if by_column = TRUE

Details

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

Value

giotto object

Examples

addGeneMetadata(gobject)

addGeneStatistics

addGeneStatistics

Description

adds gene statistics to the giotto object

Usage

```
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)
```

addHMRF addHMRF

Description

Add selected results from doHMRF to the giotto object

Usage

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

Arguments

gobject giotto object

HMRF output from doHMRF()

k number of domains

betas_to_add results from different betas that you want to add

name specify a custom name

Details

Description ...

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Value

giotto object

Examples

addHMRF(gobject)

 ${\tt addImage}$

addImage

Description

Adds giotto image objects to your giotto object

Usage

```
addImage(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

Examples

```
addImage(mg_object)
```

add Image To Spat Plot

addImageToSpatPlot

Description

Add a giotto image to a spatial ggplot object post creation

Usage

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

Arguments

spatpl a spatial ggplot object

gimage a giotto image, see createGiottoImage

Value

an updated spatial ggplot object

addNetworkLayout 15

Examples

```
addImageToSpatPlot(mg_object)
```

addNetworkLayout

addNetworkLayout

Description

Add a network layout for a selected nearest neighbor network

Usage

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

Arguments

Details

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

Value

giotto object with updated layout for selected NN network

Examples

```
addNetworkLayout(gobject)
```

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addStatistics

addStatistics

Description

adds genes and cells statistics to the giotto object

Usage

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

Arguments

Details

See addGeneStatistics and addCellStatistics

Value

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

Examples

```
addStatistics(gobject)
```

adjustGiottoMatrix adjustGiottoMatrix

Description

normalize and/or scale expresion values of Giotto object

aes_string2

Usage

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

Arguments

```
gobject giotto object

expression_values

expression values to use

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

covariate_columns

metadata columns that represent covariates to regress out

return_gobject boolean: return giotto object (default = TRUE)

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

Details

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

Value

giotto object

Examples

```
adjustGiottoMatrix(gobject)
```

```
aes_string2 aes_string2
```

Description

makes sure aes_string can also be used with names that start with numeric values

Usage

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

```
all\_plots\_save\_function \\ all\_plots\_save\_function
```

Description

Function to automatically save plots to directory of interest

Usage

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

Arguments

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

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units units

dpi Plot resolution

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

to prevent the common error of specifying dimensions in pixels.

... additional parameters to ggplot_save_function or general_save_function

See Also

```
general_save_function
```

Examples

```
all_plots_save_function(gobject)
```

annotateGiotto

annotateGiotto

Description

Converts cluster results into provided annotation.

Usage

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

Arguments

Details

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

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

Value

giotto object

Examples

```
annotateGiotto(gobject)
```

```
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

Usage

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

Arguments

spatloc spatial_locs slot from giotto object

spatgrid selected spatial_grid slot from giotto object

Value

annotated spatial location data.table

Examples

```
annotate_spatlocs_with_spatgrid_2D()
```

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

Description

annotate spatial locations with 3D spatial grid information

Usage

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

Arguments

spatloc spatial_locs slot from giotto object

spatgrid selected spatial_grid slot from giotto object

Value

annotated spatial location data.table

Examples

```
annotate_spatlocs_with_spatgrid_3D()
```

binSpect binSpect

```
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

Usage

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

calculateHVG 25

Arguments

gobject giotto object

expression_values

expression values to use

method method to calculate highly variable genes

reverse_log_scale

reverse log-scale of expression values (default = FALSE)

logbase if reverse_log_scale is TRUE, which log base was used?

expression_threshold

expression threshold to consider a gene detected

nr_expression_groups

number of expression groups for cov_groups

zscore_threshold

zscore to select hvg for cov_groups

HVGname name for highly variable genes in cell metadata

difference_in_cov

minimum difference in coefficient of variance required

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

return_gobject boolean: return giotto object (default = TRUE)

Details

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

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

2. high COV based on loess regression prediction:

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

Value

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

Examples

calculateHVG(gobject)

26 calculateMetaTableCells

calculateMetaTable

calculateMetaTable

Description

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

Usage

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

Arguments

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

Value

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

Examples

```
calculateMetaTable(gobject)
```

calculateMetaTableCells

calculate Meta Table Cells

Description

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

Usage

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

Arguments

```
gobject giotto object

value_cols metadata or enrichment value columns to use

metadata_cols annotation columns found in pDataDT(gobject)

spat_enr_names which spatial enrichment results to include
```

Value

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

Examples

```
calculateMetaTableCells(gobject)
```

```
calculate\_distance\_and\_weight \\ calculate\_distance\_and\_weight
```

Description

```
calculate_distance_and_weight
```

Usage

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

```
cell {\tt ProximityBarplot} \quad \textit{cellProximityBarplot}
```

Description

Create barplot from cell-cell proximity scores

28 cellProximityBarplot

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

giotto object gobject **CPscore** CPscore, output from cellProximityEnrichment() min_orig_ints filter on minimum original cell-cell interactions min_sim_ints filter on minimum simulated cell-cell interactions p-value p_val 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 ${\tt default_save_name}$ default save name for saving, don't change, change save_name in save_param

Details

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

Value

ggplot barplot

Examples

```
cellProximityBarplot(CPscore)
```

```
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.

Examples

```
cellProximityEnrichment(gobject)
```

```
cellProximityHeatmap cellProximityHeatmap
```

Description

Create heatmap from cell-cell proximity scores

Usage

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

Arguments

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

Details

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

Value

```
ggplot heatmap
```

cellProximityNetwork 31

Examples

```
cellProximityHeatmap(CPscore)
```

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

Description

Create network from cell-cell proximity scores

Usage

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

Arguments

```
gobject
                  giotto object
CPscore
                  CPscore, output from cellProximityEnrichment()
remove_self_edges
                  remove enrichment/depletion edges with itself
self_loop_strength
                  size of self-loops
color_depletion
                  color for depleted cell-cell interactions
color_enrichment
                  color for enriched cell-cell interactions
rescale_edge_weights
                  rescale edge weights (boolean)
{\tt edge\_weight\_range\_depletion}
                  numerical vector of length 2 to rescale depleted edge weights
```

```
edge_weight_range_enrichment
                 numerical vector of length 2 to rescale enriched edge weights
                 layout algorithm to use to draw nodes and edges
only_show_enrichment_edges
                 show only the enriched pairwise scores
edge_width_range
                 range of edge width
node_size
                 size of nodes
node_text_size size of node labels
show_plot
                 show plot
return_plot
                 return ggplot object
                 directly save the plot [boolean]
save_plot
save_param
                 list of saving parameters from all_plots_save_function
```

Details

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

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

Value

igraph plot

default_save_name

Examples

```
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 plots show_plot return_plot return ggplot object directly save the plot [boolean] save_plot list of saving parameters from all_plots_save_function save_param default_save_name default save name for saving, don't change, change save_name in save_param

Details

Description of parameters.

Value

ggplot

See Also

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

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

Details

Description of parameters.

Value

ggplot

Examples

```
cellProximitySpatPlot2D(gobject)
```

```
cell Proximity SpatPlot 3D \\ cell Proximity SpatPlot 2D
```

Description

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

Usage

```
cellProximitySpatPlot3D(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = "sdimx",
  sdimy = "sdimy",
  sdimz = "sdimz",
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = T,
  show_network = T,
  show\_other\_network = F,
  network_color = NULL,
  spatial_network_name = "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

```
cell Proximity SpatPlot 3D (gobject) \\
```

38 cellProximityVisPlot

```
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"),
)
```

cellProximityVisPlot 39

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 underlying spatial network show_network color of spatial network network_color spatial_network_name name of spatial network to use show_grid show spatial grid color of spatial grid grid_color 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

sdimx

size of other points

point_other_border_col

border color of other points

point_other_border_stroke

stroke size of other points

Details

Description of parameters.

Value

ggplot or plotly

Examples

cellProximityVisPlot(gobject)

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

Description

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

Usage

```
cellProximityVisPlot_2D_ggplot(
  gobject,
  interaction_name = NULL,
  cluster_column = NULL,
  sdimx = NULL,
  sdimy = NULL,
  cell_color = NULL,
  cell_color_code = NULL,
  color_as_factor = T,
  show_other_cells = F,
  show_network = F,
  show_other_network = F,
  network_color = NULL,
  spatial_network_name = "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

named vector with giotto instructions

Examples

changeGiottoInstructions()

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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")
```

Arguments

mg_object magick image object

bg_color estimated current background color

perc_range range around estimated background color to include (percentage)

new_color new background color

Value

vector of pixel color frequencies and an associated barplot

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,
  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,
```

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```
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
   pyth\_leid\_resolution
                    resolution for leiden
   pyth_leid_weight_col
                    column to use for weights
   pyth_leid_part_type
                    partition type to use
   pyth_leid_init_memb
                    initial membership
   pyth_leid_iterations
```

number of iterations

48 clusterCells

pyth_louv_resolution

resolution for louvain

pyth_louv_weight_col

python louvain param: weight column

python_louv_random

python louvain param: random

python_path specify specific path to python if required

louvain_gamma louvain param: gamma or resolution

louvain_omega louvain param: omega

walk_steps randomwalk: number of steps
walk_clusters randomwalk: number of clusters
walk_weights randomwalk: weight column
sNNclust_k SNNclust: k neighbors to use

sNNclust_eps SNNclust: epsilon

sNNclust_minPts

SNNclust: min points

borderPoints SNNclust: border points

expression_values

expression values to use

genes_to_use = NULL,
dim_reduction_to_use

dimension reduction to use

dim_reduction_name

name of reduction 'pca',

dimensions_to_use

dimensions to use

distance_method

distance method

km_centers kmeans centers km_iter_max kmeans iterations

km_nstart kmeans random starting points

km_algorithm kmeans algorithm

hc_agglomeration_method

hierarchical clustering method

hc_k hierachical number of clusters

hc_h hierarchical tree cutoff

return_gobject boolean: return giotto object (default = TRUE)

set_seed set seed

seed_number number for seed

Details

Wrapper for the different clustering methods.

Value

giotto object with new clusters appended to cell metadata

See Also

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

Examples

```
clusterCells(gobject)
```

clusterSpatialCorGenes

clusterSpatialCorGenes

Description

Cluster based on spatially correlated genes

Usage

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

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 colMeans_giotto

Description

```
colMeans_giotto
```

Usage

```
colMeans_giotto(mymatrix)
```

50 combCCcom

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
)
```

Arguments

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

Value

combined data.table with spatial and expression communication data

```
combCCcom(gobject)
```

```
combineCellProximityGenes
```

combineCellProximityGenes

Description

Combine CPG scores in a pairwise manner.

Usage

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

Arguments

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

Value

cpgObject that contains the filtered differential gene scores

52 combineCPG

Examples

```
combineCellProximityGenes(gobject)
```

```
combine Cell Proximity Genes\_per\_interaction \\ combine Cell Proximity Genes\_per\_interaction
```

Description

Combine CPG scores per interaction

Usage

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

Examples

combineCellProximityGenes_per_interaction()

combineCPG

combineCPG

Description

Combine CPG scores in a pairwise manner.

Usage

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

combineMetadata 53

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

Arguments

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

Value

cpgObject that contains the filtered differential gene scores

Examples

```
combineCPG(gobject)
```

combineMetadata combineMetadata

Description

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

Usage

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

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

Value

Extended cell metadata in data.table format.

Examples

```
combineMetadata(gobject)
```

```
{\tt convertEnsemblToGeneSymbol}
```

convertEnsemblToGeneSymbol

Description

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

Usage

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

Arguments

matrix an expression matrix with ensembl gene IDs as rownames

species species to use for gene symbol conversion

Details

This function requires that the biomaRt library is installed

Value

expression matrix with gene symbols as rownames

Examples

```
convertEnsemblToGeneSymbol(matrix)
```

```
{\it convert\_mgImage\_to\_array\_DT} \\ {\it convert\_mgImage\_to\_array\_DT}
```

Description

converts a magick image object to a data.table

Usage

```
convert_mgImage_to_array_DT(mg_object)
```

Arguments

```
mg_object magick 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)
```

createCrossSection createCrossSection

Description

Create a virtual 2D cross section.

56 createCrossSection

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
)
```

Arguments

gobject giotto object

name name of cress section object. (default = cross_section)

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)

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

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

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

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

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

createGiottoImage 57

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 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
)
```

```
giotto object
gobject
spatial\_locs
                  spatial locations
mg_object
                  magick image object
                  name for the image
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
                  adjustment of the maximum y-value to align the image
ymax_adj
ymin_adj
                  adjustment of the minimum y-value to align the image
```

58 createGiottoInstructions

Value

```
a giotto image object
```

Examples

```
createGiottoImage(mg_object)
```

createGiottoInstructions

create Giot to Instructions

Description

Function to set global instructions for giotto functions

Usage

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

Arguments

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

Value

named vector with giotto instructions

```
createGiottoInstructions()
```

createGiottoObject 59

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)
```

60 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)

createHeatmap_DT 61

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"
)
```

Arguments

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

Details

Creates input data.tables for plotHeatmap function.

62 createMetagenes

Value

list

Examples

```
createHeatmap_DT(gobject)
```

createMetagenes

createMetagenes

Description

This function creates an average metagene for gene clusters.

Usage

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

Arguments

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

Details

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

Value

giotto object

```
createMetagenes(gobject)
```

createNearestNetwork 63

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)

 $create {\tt Spatial Delaunay Network} \\ create {\tt Spatial Delaunay Network}$

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

```
createSpatialDelaunayNetwork(gobject)
```

66 createSpatialEnrich

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
```

createSpatialGrid 67

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.

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

```
createSpatialGrid_2D(gobject)
```

createSpatialGrid_3D 69

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

Description

Create a spatial grid for 3D spatial data.

Usage

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

Arguments

```
gobject giotto object

sdimx_stepsize stepsize along the x-axis

sdimy_stepsize stepsize along the y-axis

sdimz_stepsize stepsize along the z-axis

minimum_padding

minimum padding on the edges

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

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

Details

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

Value

giotto object with updated spatial grid slot

```
createSpatialGrid_3D(gobject)
```

```
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)

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

```
createSpatialKNNnetwork(gobject)
```

createSpatialNetwork 71

```
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("delaunayn_geometry", "RTriangle", "deldir"),
  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,
)
```

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_distand	ce_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

Usage

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

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

Description

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

Usage

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

Arguments

```
gobject giotto object

meta_data_name name of metadata column to use
expression_values

which expression values to use

detection_threshold

detection threshold to consider a gene detected
```

Value

data.table with average gene epression values for each factor

Description

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

Usage

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

Arguments

```
gobject giotto object

meta_data_name name of metadata column to use
expression_values

which expression values to use
```

Value

data.table with average gene epression values for each factor

Description

creates randomized cell ids within a selection of cell types

Usage

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

Arguments

```
gobject giotto object to use

cluster_column cluster column with cell type information

needed_cell_types

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

Details

Details will follow.

Value

list of randomly sampled cell ids with same cell type composition

Examples

```
create_cell_type_random_cell_IDs(gobject)
```

create_cluster_matrix 75

```
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 79

```
create_dimObject
```

Description

Creates an object that stores a dimension reduction output

Usage

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

Arguments

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

Value

number of distinct colors

```
create_jackstrawplot create_jackstrawplot
```

Description

create jackstrawplot with ggplot

```
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)
```

Description

create mesh grid lines for cross section

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

create_screeplot 81

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

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

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crossSectionGenePlot crossSectionGenePlot

Description

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

Usage

```
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,
  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 name of virtual cross section to use
```

spatial_network_name name of spatial network to use expression_values gene expression values to use genes to show genes genes_high_color color represents high gene expression genes_mid_color color represents middle gene expression genes_low_color color represents low gene expression show_network show underlying spatial network 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 point_border_stroke stroke size of border around points show_legend show legend legend_text size of legend text background_color color of plot background size of axis text axis_text axis_title size of axis title cow_n_col cowplot param: how many columns cowplot param: relative height cow_rel_h 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] 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

See Also

```
spatGenePlot3D and spatGenePlot2D
```

Examples

```
crossSectionGenePlot(gobject)
```

crossSectionGenePlot3D

crossSectionGenePlot3D

Description

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

```
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,
```

crossSectionGenePlot3D 85

```
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 underlying spatial network
    show_network
    network_color
                     color of spatial network
                     show spatial grid
    show_grid
    genes_high_color
                     color represents high gene expression
    genes_mid_color
                     color represents middle gene expression
    genes_low_color
                     color represents low gene expression
    spatial_grid_name
                     name of spatial grid to use
    point_size
                     size of point (cell)
                     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
                     color of spatial grid
    grid_color
    midpoint
                     expression midpoint
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
                     parameters for cowplot::save_plot()
```

Details

Description of parameters.

x_ticks = NULL, y_ticks = NULL, 86 crossSectionPlot

Value

ggplot

Examples

crossSectionGenePlot3D(gobject)

crossSectionPlot

crossSectionPlot

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",
```

crossSectionPlot 87

```
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 subset of cells based on cell IDs
    select_cells
```

point with border or not (border or no_border)

point_shape

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size of point (cell) point_size point_border_col color of border around points point_border_stroke stroke size of border around points show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels show_network show underlying spatial network network_color color of spatial network network_alpha alpha of spatial network show_grid show spatial grid spatial_grid_name name of spatial grid to use grid_color color of spatial grid show_other_cells display not selected cells other_cell_color color of not selected cells other_point_size point size of not selected cells other_cells_alpha alpha of not selected cells coord_fix_ratio fix ratio between x and y-axis title title of plot show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols background_color color of plot background axis_text size of axis text axis_title size of axis title cowplot param: how many columns cow_n_col cowplot param: relative height cow_rel_h cow_rel_w cowplot param: relative width cowplot param: how to align cow_align

show_plot

show plot

crossSectionPlot3D 89

Details

Description of parameters.

Value

ggplot

See Also

crossSectionPlot

crossSectionPlot3D

Description

Visualize cells in a virtual cross section according to spatial coordinates

```
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 = "",
```

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```
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 of virtual cross section to use
    name
    spatial_network_name
                      name of spatial network to use
                      x-axis dimension name (default = 'sdimx')
    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 of plot
    title
    show_legend
                      show legend
    axis_scale
                      the way to scale the axis
                      customize the scale of the plot
    custom_ratio
    x_ticks
                      set the number of ticks on the x-axis
                      set the number of ticks on the y-axis
    y_ticks
                      set the number of ticks on the z-axis
    z_ticks
```

show_plot

show plot

decide_cluster_order 91

Details

Description of parameters.

Value

ggplot

Examples

```
crossSectionPlot3D(gobject)
```

```
decide_cluster_order decide_cluster_order
```

Description

creates order for clusters

Usage

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

Arguments

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

Details

Calculates order for clusters.

Value

custom

Examples

```
decide_cluster_order(gobject)
```

 ${\tt detectSpatialCorGenes} \ \ \textit{detectSpatialCorGenes}$

Description

Detect genes that are spatially correlated

Usage

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

Arguments

```
giotto object
gobject
method
                  method to use for spatial averaging
expression_values
                  gene expression values to use
                  subset of genes to use
subset_genes
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)
```

detectSpatialPatterns 93

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()

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.

```
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
)
```

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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_unit
                  scale features
                  number of principal components to calculate
ncp
show_plot
                  show plots
PC_zscore
                  minimum z-score of variance explained by a PC
```

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",
```

dimCellPlot 95

```
cell_color_gradient = c("blue", "white", "red"),
      gradient_midpoint = NULL,
      gradient_limits = NULL,
      select_cell_groups = NULL,
      select_cells = NULL,
      show_other_cells = T,
     other_cell_color = "lightgrey",
     other_point_size = 0.5,
      show_cluster_center = F,
      show_center_label = T,
     center_point_size = 4,
      center_point_border_col = "black",
      center_point_border_stroke = 0.1,
      label_size = 4,
      label_fontface = "bold",
      edge_alpha = NULL,
     point_shape = c("border", "no_border"),
     point_size = 1,
     point_border_col = "black",
     point_border_stroke = 0.1,
      show_legend = T,
     legend_text = 8,
      legend_symbol_size = 1,
     background_color = "white",
      axis_text = 8,
     axis_title = 8,
      cow_n_col = 2,
      cow_rel_h = 1,
      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
                    dimension to use on y-axis
   dim2_to_use
   spat_enr_names names of spatial enrichment results to include
   cell_annotation_values
```

numeric cell annotation columns

show underlying NN network

show_NN_network

network_name = "sNN.pca",

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```
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
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
point_shape
                  point with border or not (border or no_border)
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
axis_text
                  size of axis text
                  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
```

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```
default_save_name

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

cell_color color for cells (see details)

color_as_factor

convert color column to factor

cell_color_code

named vector with colors

title title for plot, defaults to cell_color parameter
```

Details

Description of parameters. For 3D plots see dimCellPlot2D

Value

ggplot

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,
```

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```
center_point_size = 4,
      center_point_border_col = "black",
      center_point_border_stroke = 0.1,
      label_size = 4,
      label_fontface = "bold",
      edge_alpha = NULL,
     point_shape = c("border", "no_border"),
      point_size = 1,
     point_border_col = "black",
     point_border_stroke = 0.1,
      show_legend = T,
     legend_text = 8,
      legend_symbol_size = 1,
     background_color = "white",
     axis_text = 8,
      axis_title = 8,
      cow_n_col = 2,
     cow_rel_h = 1,
     cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
     return_plot = NA,
      save_plot = NA,
     save_param = list(),
     default_save_name = "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
                    dimension to use on y-axis
   dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                    numeric cell annotation columns
    show_NN_network
                    show underlying NN network
   nn_network_to_use
                    type of NN network to use (kNN vs sNN)
   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 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_shape point with border or not (border or no_border) point_size size of point (cell) point_border_col color of border around points point_border_stroke stroke size of border around points show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols background_color color of plot background axis_text size of axis text axis_title size of axis title 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

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Details

Description of parameters. For 3D plots see dimPlot3D

Value

ggplot

Examples

```
dimCellPlot2D(gobject)
```

dimGenePlot

dimGenePlot

Description

Visualize cells and gene expression according to dimension reduction coordinates

```
dimGenePlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  genes = NULL,
  dim_reduction_to_use = "umap",
  dim_reduction_name = "umap",
  dim1_to_use = 1,
  dim2\_to\_use = 2,
  show_NN_network = F,
  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,
  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,
```

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```
return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot"
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
                     genes to show
    genes
   dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
   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
                     column to use for alpha of the edges
    edge_alpha
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
                     size of point (cell)
    point_size
    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
                     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 plots
    return_plot
                     return ggplot object
    save_plot
                     directly save the plot [boolean]
                     list of saving parameters from all_plots_save_function
    save_param
    default_save_name
                     default save name for saving, don't change, change save_name in save_param
                     parameters for cowplot::save_plot()
    . . .
```

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Details

Description of parameters.

Value

ggplot

See Also

dimGenePlot3D

Examples

```
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,
  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,
```

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```
cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "dimGenePlot2D"
Arguments
    gobject
                     giotto object
   expression_values
                     gene expression values to use
                     genes to show
    genes
    dim_reduction_to_use
                     dimension reduction to use
   dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
   dim2_to_use
                     dimension to use on y-axis
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
   network_name
    edge_alpha
                     column to use for alpha of the edges
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
                     point with border or not (border or no_border)
   point_shape
   point_size
                     size of point (cell)
    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

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```
size of axis title
axis_title
                  cowplot param: how many columns
cow_n_col
                  cowplot param: relative height
cow_rel_h
                  cowplot param: relative width
cow_rel_w
                  cowplot param: how to align
cow_align
show_plot
                  show 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",
```

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

. . .

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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_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
                     giotto object
   gobject
   group_by_subset
                     subset the group_by factor column
    dim_reduction_to_use
                     dimension reduction to use
   dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
   dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
   nn_network_to_use
                     type of NN network to use (kNN vs sNN)
   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
```

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```
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
center_point_size
                  size of center points
label_size
                  size of labels
label_fontface font of labels
                  column to use for alpha of the edges
edge_alpha
                  point with border or not (border or no_border)
point_shape
point_size
                  size of point (cell)
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
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
                  title for plot, defaults to cell_color parameter
                  cowplot param: how many columns
cow_n_col
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
                  cowplot param: how to align
cow_align
show_plot
                  show plot
return_plot
                  return ggplot object
                  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
```

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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_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
                     giotto object
   gobject
   group_by_subset
                     subset the group_by factor column
    dim_reduction_to_use
                     dimension reduction to use
   dim_reduction_name
                     dimension reduction name
                     dimension to use on x-axis
   dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
   nn_network_to_use
                     type of NN network to use (kNN vs sNN)
   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 other_point_size size of not selected cells show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels edge_alpha column to use for alpha of the edges point with border or not (border or no_border) point_shape point_size size of point (cell) point_border_col color of border around points point_border_stroke stroke size of border around points title for plot, defaults to cell_color parameter title show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols background_color color of plot background axis_text size of axis text axis_title size of axis title cow_n_col cowplot param: how many columns cow_rel_h cowplot param: relative height cow_rel_w cowplot param: relative width cowplot param: how to align cow_align show_plot show plot return_plot return ggplot object save_plot directly save the plot [boolean] list of saving parameters from all_plots_save_function save_param default_save_name default save name for saving, don't change, change save_name in save_param create multiple plots based on cell annotation column groub_by

112 dimPlot2D_single

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,
```

dimPlot2D_single 113

```
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
    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
                     dimension to use on y-axis
    dim2_to_use
    spat_enr_names names of spatial enrichment results to include
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show NN network = TRUE
    network_name
                     color for cells (see details)
    cell_color
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
    select_cells
                     select subset of cells based on cell IDs
    show_other_cells
                     display not selected cells
    other_cell_color
                     color of not selected cells
```

114 dimPlot2D_single

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

Details

Description of parameters. For 3D plots see dimPlot3D

Value

ggplot

Examples

```
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"
)
```

dimPlot3D

```
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
show\_legend
                  show legend
```

Details

Description of parameters.

Value

plotly

Examples

dimPlot3D(gobject)

doHclust 117

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
                  subset of genes to use
genes_to_use
dim_reduction_to_use
                  dimension reduction to use
dim_reduction_name
                  dimensions reduction name
dimensions_to_use
                  dimensions to use
distance_method
                  distance method
{\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
```

118 doHMRF

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 119

```
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 121

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
```

122 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 123

```
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)

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

124 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,
)
```

number for seed

Arguments

seed_number

```
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
```

Details

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

Value

giotto object with new clusters appended to cell metadata

See Also

doLouvainCluster_community and doLouvainCluster_multinet

Examples

```
doLouvainCluster(gobject)
```

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

Description

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

Usage

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

weight_col weight column to use for edges

louv_random Will randomize the node evaluation order and the community evaluation order

to get different partitions at each call

return_gobject boolean: return giotto object (default = TRUE)

set_seed set seed

seed_number number for seed

Details

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

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

Value

giotto object with new clusters appended to cell metadata

Examples

```
doLouvainCluster_community(gobject)
```

```
doLouvainCluster_multinet
```

doLouvainCluster_multinet

Description

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

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

doLouvainSubCluster 127

Arguments

Details

seed_number

See glouvain_ml from the multinet package in R for more information.

Value

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,
```

128 doLouvainSubCluster

```
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{lower} do Louvain SubCluster\_community \\ do Louvain SubCluster\_community
```

Description

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

Usage

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

```
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)

 ${\tt network_name} \quad \quad {\tt name} \ of \ NN \ network \ to \ use$

return_gobject boolean: return giotto object (default = TRUE)

verbose verbose

Details

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

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

Value

giotto object with new subclusters appended to cell metadata

See Also

doLouvainCluster_community

Examples

doLouvainSubCluster_community(gobject)

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

Description

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

Usage

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

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

132 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)

network_name name of NN network to use

return_gobject boolean: return giotto object (default = TRUE)

verbose verbose

python_path specify specific path to python if required

Details

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

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

Value

giotto object with new subclusters appended to cell metadata

See Also

doLouvainCluster_multinet

Examples

doLouvainSubCluster_multinet(gobject)

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

Description

Cluster cells using a random walk approach.

doRandomWalkCluster 133

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)
```

134 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 135

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,
    mean_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 141

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)
```

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

142 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 143

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
```

Arguments

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

Details

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

Value

writes the necessary output to use in Giotto Viewer

144 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

Examples

```
exprCellCellcom(gobject)
```

extended_gini_fun 145

```
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 147

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")
```

148 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
                  default save name for saving, don't change, change save_name in save_param
```

Details

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

Value

list of data.table and ggplot object

Examples

filterCombinations(gobject)

filterCPG filterCPG

Description

Filter cell proximity gene scores.

filterDistributions 149

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

150 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 151

filterGiotto

filterGiotto

Description

filter Giotto object based on expression threshold

Usage

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

Arguments

```
gobject giotto object

expression_values

expression values to use

expression_threshold

threshold to consider a gene expressed

gene_det_in_min_cells

minimum # of cells that need to express a gene

min_det_genes_per_cell

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

verbose

verbose
```

Details

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

Value

giotto object

```
filterGiotto(gobject)
```

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,
   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

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"),
  mean_method = c("arithmic", "geometric"),
```

findCPG 155

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
cores
                  number of cores to use if do_parallel = TRUE
```

Details

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

- genes: All or selected list of tested genes
- sel: average gene expression in the interacting cells from the target cell type
- other: average gene expression in the NOT-interacting cells from the target cell type
- log2fc: log2 fold-change between sel and other
- · diff: spatial expression difference between sel and other
- p.value: associated p-value
- p.adj: adjusted p-value
- cell_type: target cell type

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- 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

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

cluster_column clusters to use
subset_clusters
selection of clusters to compare
group_1 group 1 cluster IDs from cluster_column for pairwise comparison
group_2 group 2 cluster IDs from cluster_column for pairwise comparison
```

```
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 https://en.wikipedia.org/wiki/Gini_coefficientgini 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 159

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

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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

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

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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)
```

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```
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)

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)
```

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

Description

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

```
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)

 $\begin{tabular}{ll} find_grid_2D & find_grid_2D \end{tabular}$

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

Usage

find_x_y_ranges(data, extend_ratio)

```
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)
```

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get10Xmatrix

get10Xmatrix

Description

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

Usage

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

Arguments

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

Details

A typical 10X folder is named raw_feature_bc_matrix or raw_feature_bc_matrix and tt has 3 files:

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

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

Value

expression matrix from 10X

Examples

```
get10Xmatrix(10Xmatrix)
```

```
{\tt getClusterSimilarity} \quad \textit{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")
)
```

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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
```

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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

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

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

Description

get local coordinates within cross section plane

```
get_cross_section_coordinates(cell_subset_projection_locations)
```

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get_distance

get_distance

Description

estimate average distance between neighboring cells with network table as input

Usage

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

```
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
)
```

```
{\tt ggplot\_save\_function} \quad {\tt ggplot\_save\_function}
```

Description

Function to automatically save plots to directory of interest

```
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,
```

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to prevent the common error of specifying dimensions in pixels.

```
nrow = 1,
scale = 1,
base_width = NULL,
base_height = NULL,
base_aspect_ratio = NULL,
units = NULL,
dpi = NULL,
limitsize = TRUE,
...
)
```

Arguments

gobject giotto object ggplot object to plot plot_object save_dir directory to save to folder in save_dir to save to save_folder save_name name of plot save_format format (e.g. png, tiff, pdf, ...) show_saved_plot load & display the saved plot number of columns ncol number of rows nrow scale scale base_width width base_height height base_aspect_ratio aspect ratio units units dpi Plot resolution When TRUE (the default), ggsave will not save images larger than 50x50 inches,

See Also

```
cowplot::save_plot
```

```
{\tt ggplot\_save\_function(gobject)}
```

giotto-class

S4 giotto Class

Description

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

Slots

```
raw_exprs raw expression counts
norm_expr normalized expression counts
norm_scaled_expr normalized and scaled expression counts
custom_expr custom normalized counts
spatial_locs spatial location coordinates for cells
cell_metadata metadata for cells
gene_metadata metadata for genes
cell_ID unique cell IDs
gene_ID unique gene IDs
spatial_network spatial network in data.table/data.frame format
spatial_grid spatial grid in data.table/data.frame format
dimension_reduction slot to save dimension reduction coordinates
nn_network nearest neighbor network in igraph format
parameters slot to save parameters that have been used
instructions slot for global function instructions
offset_file offset file used to stitch together image fields
OS_platform Operating System to run Giotto analysis on
```

 $heatmSpatialCorGenes \quad \textit{heatmSpatialCorGenes}$

Description

Create heatmap of spatially correlated genes

```
heatmSpatialCorGenes(
  gobject,
  spatCorObject,
  use_clus_name = NULL,
  show_cluster_annot = TRUE,
  show_row_dend = T,
  show_column_dend = F,
  show_row_names = F,
```

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```
show_column_names = F,
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "heatmSpatialCorGenes",
...
)
```

Arguments

```
giotto object
gobject
                 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 dendrogram
show_row_dend
show_column_dend
                 show column dendrogram
show_row_names show row names
show_column_names
                 show column names
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
                 additional parameters to the Heatmap function from ComplexHeatmap
. . .
```

Value

Heatmap generated by ComplexHeatmap

Examples

```
heatmSpatialCorGenes(gobject)
```

 $hyperGeometric Enrich \qquad hyperGeometric Enrich$

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)
```

```
insert {\tt CrossSectionGenePlot3D}
```

insert Cross Section Gene Plot 3D

Description

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

Usage

```
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

```
gobject giotto object

name 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
                  x-axis dimension name (default = 'sdimx')
sdimx
                  y-axis dimension name (default = 'sdimy')
sdimy
sdimz
                  z-axis dimension name (default = 'sdimy')
expression_values
                  gene expression values to use
                  genes to show
genes
show_network
                  show underlying spatial network
                  color of spatial network
network_color
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
                  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
midpoint
                  expression midpoint
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  parameters for cowplot::save_plot()
```

Details

Description of parameters.

Value

ggplot

Examples

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 x-axis dimension name (default = 'sdimx') sdimx y-axis dimension name (default = 'sdimy') sdimy sdimz z-axis dimension name (default = 'sdimy') point_size size of point (cell) cell_color color for cells (see details) cell_color_code named vector with colors select_cell_groups select subset of cells/clusters based on cell_color parameter 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 color of spatial grid grid_color 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 set the number of ticks on the x-axis x_ticks y_ticks set the number of ticks on the y-axis set the number of ticks on the z-axis z_ticks 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 jackstrawPlot 181

Details

Description of parameters.

Value

ggplot

Examples

insertCrossSectionSpatPlot3D(gobject)

jackstrawPlot

jackstrawPlot

Description

identify significant prinicipal components (PCs)

Usage

```
jackstrawPlot(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  genes_to_use = NULL,
  center = F,
  scale_unit = F,
  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"
)
```

```
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
```

182 libNorm_giotto

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. multiple PCA results can be stored by changing the *name* parameter

Value

ggplot object for jackstraw method

Examples

jackstrawPlot(gobject)

Description

create binarized scores from a vector using kmeans

Usage

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

 ${\tt libNorm_giotto} \qquad \qquad {\it libNorm_giotto}$

Description

libNorm_giotto

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

loadHMRF 183

loadHMRF

loadHMRF

Description

load previous HMRF

Usage

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

Arguments

Details

Description of HMRF parameters ...

Value

reloads a previous ran HMRF from doHRMF

Examples

```
loadHMRF(gobject)
```

logNorm_giotto

logNorm_giotto

Description

```
logNorm_giotto
```

```
logNorm_giotto(mymatrix, base, offset)
```

184 makeSignMatrixRank

makeSignMatrixPAGE makeSignMatrixPAGE

Description

Function to convert a list of signature genes (e.g. for cell types or processes) into a binary matrix format that can be used with the PAGE enrichment option. Each cell type or process should have a vector of cell-type or process specific genes. These vectors need to be combined into a list (sign_list). The names of the cell types or processes that are provided in the list need to be given (sign_names).

Usage

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

Arguments

sign_names vector with names for each provided gene signature

sign_list list of genes (signature)

Value

matrix

See Also

PAGEEnrich

Examples

makeSignMatrixPAGE()

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

Description

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

Usage

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

Arguments

sc_matrix matrix of single-cell RNAseq expression data

sc_cluster_ids vector of cluster ids

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

ered

Value

matrix

See Also

rankEnrich

Examples

```
makeSignMatrixRank()
```

```
make_simulated_network
```

 $make_simulated_network$

Description

Simulate random network.

Usage

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

Examples

```
make_simulated_network(gobject)
```

```
mean_expr_det_test
```

mean_expr_det_test

Description

```
mean_expr_det_test
```

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

186 mergeClusters

mergeClusters

mergeClusters

Description

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

Usage

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

Arguments

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

mygini_fun 187

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

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)
```

node_clusters

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)
```

Examples

```
my_rowMeans(x)
```

nnDT_to_kNN

nnDT_to_kNN

Description

Convert a nearest network data.table to a kNN object

Usage

```
nnDT_to_kNN(nnDT)
```

Arguments

nnDT

nearest neighbor network in data.table format

Value

kNN object

 ${\tt node_clusters}$

 $node_clusters$

Description

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

Usage

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

Arguments

hclus_obj hclus object verbose be verbose normalizeGiotto 189

Value

list of splitted dendrogram nodes from high to low node height

Examples

```
node_clusters(hclus_obj)
```

normalize Giotto

normalize Giotto

Description

fast normalize and/or scale expresion values of Giotto object

Usage

```
normalizeGiotto(
  gobject,
  norm_methods = c("standard", "osmFISH"),
  library_size_norm = TRUE,
  scalefactor = 6000,
  log_norm = TRUE,
  log_offset = 1,
  logbase = 2,
  scale_genes = T,
  scale_cells = T,
  scale_order = c("first_genes", "first_cells"),
  verbose = F
```

```
gobject
                  giotto object
                  normalization method to use
norm_methods
library_size_norm
                  normalize cells by library size
                  scale factor to use after library size normalization
scalefactor
                  transform values to log-scale
log_norm
                  offset value to add to expression matrix, default = 1
log_offset
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
```

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Details

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

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

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

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

Value

giotto object

Examples

normalizeGiotto(gobject)

PAGEEnrich

PAGEEnrich

Description

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

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

pca_giotto 191

Arguments

gobject Giotto object

sign_matrix Matrix of signature genes for each cell type / process

expression_values

expression values to use

reverse_log_scale

reverse expression values from log scale

logbase log base to use if reverse_log_scale = TRUE

output_enrichment

how to return enrichment output

Details

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

The enrichment Z score is calculated by using method (PAGE) from Kim SY et al., BMC bioinformatics, 2005 as $Z=((Sm\check{\ }mu)*m^(1/2))/delta$. 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

192 plotCCcomDotplot

Value

list of eigenvalues, eigenvectors and pca coordinates

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 193

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

196 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"
)
```

```
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

Examples

```
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 201

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"
)
```

```
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
```

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```
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
cell_color_code
                  vector of colors with cell types as names
show_plot
                  show plots
                  return plotting object
return_plot
                  directly save the plot [boolean]
save_plot
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)
```

plotHeatmap

plotHeatmap

Description

Creates heatmap for genes and clusters.

```
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,
```

plotHeatmap 203

```
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 to use
    genes
    cluster_column name of column to use for clusters
                     method to determine cluster order
    cluster_order
    cluster_custom_order
                     custom order for clusters
    cluster_color_code
                     color code for clusters
    cluster_cor_method
                     method for cluster correlation
    cluster_hclust_method
                     method for hierarchical clustering of clusters
                     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
    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
                     number of rows for the cluster legend
    legend_nrows
                     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

204 plotICG

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

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"
)
```

```
gobject
                  giotto object
                  cell proximity gene score object
cpgObject
                  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 plots
show_plot
return_plot
                  return plotting object
save_plot
                  directly save the plot [boolean]
```

```
save\_param \qquad 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

```
plotICG(CPGscores)
```

```
\verb|plotInteractionChangedGenes||
```

plotInteractionChangedGenes

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")
```

```
gobject
                  giotto object
cpgObject
                  cell proximity gene score object
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
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
```

plotly_axis_scale_2D

Value

plot

Examples

```
plotInteractionChangedGenes(CPGscores)
```

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

Description

adjust the axis scale in 3D plotly plot

Usage

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

Arguments

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

Value

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

Examples

```
plotly_axis_scale_2D(gobject)
```

plotly_axis_scale_3D 207

```
plotly_axis_scale_3D plotly_axis_scale_3D
```

Description

adjust the axis scale in 3D plotly plot

Usage

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

Arguments

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

Value

edges in spatial grid as data.table()

Examples

```
plotly_axis_scale_3D(gobject)
```

```
plotly_grid
```

 $plotly_grid$

Description

provide grid segment to draw in plot_ly()

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

208 plotly_network

Arguments

```
spatial_grid spatial_grid in giotto object
```

Value

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

Examples

```
plotly_grid(gobject)
```

plotly_network

 $plotly_network$

Description

```
provide network segment to draw in 3D plot_ly()
```

Usage

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

Arguments

```
gobject network in giotto object
```

Value

```
edges in network as data.table()
```

Examples

```
plotly_network(gobject)
```

Description

Creates heatmap for numeric cell metadata within aggregated clusters.

Usage

```
plotMetaDataCellsHeatmap(
  gobject,
  metadata_cols = NULL,
  spat_enr_names = NULL,
  value_cols = NULL,
  first_meta_col = NULL,
  second_meta_col = NULL,
  show_values = c("zscores", "original", "zscores_rescaled"),
  custom_cluster_order = NULL,
  clus_cor_method = "pearson",
  clus_cluster_method = "complete",
  custom_values_order = NULL,
  values_cor_method = "pearson",
  values_cluster_method = "complete",
  midpoint = 0,
  x_{text_size} = 8,
  x_{text_angle} = 45,
  y_{text_size} = 8,
  strip_text_size = 8,
  show_plot = NA,
  return_plot = NA,
  save_plot = NA,
  save_param = list(),
  default_save_name = "plotMetaDataCellsHeatmap"
)
```

```
gobject giotto object

metadata_cols annotation columns found in pDataDT(gobject)

spat_enr_names spatial enrichment results to include

value_cols value columns to use

first_meta_col if more than 1 metadata column, select the x-axis factor

second_meta_col

if more than 1 metadata column, select the facetting factor

show_values which values to show on heatmap

custom_cluster_order

custom cluster order (default = NULL)
```

```
clus_cor_method
                  correlation method for clusters
clus_cluster_method
                  hierarchical cluster method for the clusters
                  midpoint of show_values
midpoint
                  size of x-axis text
x_text_size
x_text_angle
                  angle of x-axis text
y_text_size
                  size of y-axis text
strip_text_size
                  size of strip text
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
custom_gene_order
                  custom gene order (default = NULL)
gene_cor_method
                  correlation method for genes
gene_cluster_method
                  hierarchical cluster method for the genes
```

Details

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

Value

ggplot or data.table

See Also

plotMetaDataHeatmap for gene expression instead of numeric cell annotation data.

Examples

plotMetaDataCellsHeatmap(gobject)

plotMetaDataHeatmap 211

plotMetaDataHeatmap plotMetaDataHeatmap

Description

Creates heatmap for genes within aggregated clusters.

Usage

```
plotMetaDataHeatmap(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  metadata_cols = NULL,
  selected_genes = NULL,
  first_meta_col = NULL,
  second_meta_col = NULL,
  show_values = c("zscores", "original", "zscores_rescaled"),
  custom_cluster_order = NULL,
  clus_cor_method = "pearson",
  clus_cluster_method = "complete",
  custom_gene_order = NULL,
  gene_cor_method = "pearson",
  gene_cluster_method = "complete",
  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"
```

```
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
x_text_size
                  size of x-axis text
                  angle of x-axis text
x_text_angle
y_text_size
                  size of y-axis text
strip_text_size
                  size of strip text
show_plot
                  show plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name
```

Details

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

Value

ggplot or data.table

See Also

plotMetaDataCellsHeatmap for numeric cell annotation instead of gene expression.

Examples

plotMetaDataHeatmap(gobject)

plotPCA 213

plotPCA plotPCA

Description

Short wrapper for PCA visualization

Usage

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

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

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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_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_legend
                  show legend
title
                  title for plot, defaults to cell_color parameter
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
background_color
                  color of plot background
                  size of axis text
axis_text
                  size of axis title
axis_title
                  cowplot param: how many columns
cow_n_col
                  cowplot param: relative height
cow_rel_h
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
```

Details

Description of parameters, see dimPlot2D. For 3D plots see plotPCA_3D

Value

ggplot

Examples

plotPCA(gobject)

plotPCA_2D 215

plotPCA_2D plotPCA_2D

Description

Short wrapper for PCA visualization

Usage

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

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

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

Details

Description of parameters, see dimPlot2D. For 3D plots see plotPCA_3D

Value

ggplot

plotPCA_3D 217

Examples

```
plotPCA_2D(gobject)
```

plotPCA_3D

plotPCA_3D

Description

Visualize cells according to 3D PCA dimension reduction

Usage

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

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

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

Details

Description of parameters.

Value

plotly

Examples

```
plotPCA_3D(gobject)
```

plotRankSpatvsExpr plotRankSpatvsExpr

Description

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

```
plotRankSpatvsExpr(
  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,
```

plotRecovery 219

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

Arguments

gobject giotto object

combCC combined communinication scores from combCCcom

expr_rnk_column

column with expression rank information to use

spat_rnk_column

column with spatial rank information to use

midpoint midpoint of colors size_range size ranges of dotplot

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

return_plot return plotting object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

Value

ggplot

Examples

plotRankSpatvsExpr(CPGscores)

plotRecovery

plotRecovery

Description

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

220 plotRecovery_sub

Usage

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

Arguments

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

```
plotRecovery(CPGscores)
```

```
plotRecovery_sub plotRecovery_sub
```

Description

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

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

Arguments

```
combCC combined communication scores from combCCcom
first_col first column to use
second_col second column to use
```

Examples

```
plotRecovery_sub(CPGscores)
```

```
\verb"plotStatDelaunayNetwork"
```

plotStatDelaunayNetwork

Description

Plots network statistics for a Delaunay network..

Usage

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

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

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

show_plot show plots

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

Other parameters of the triangulate function

name for spatial network (default = 'delaunay_network')

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 plotTSNE

Description

Short wrapper for tSNE visualization

Usage

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

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 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_border_col color of border around points point_border_stroke stroke size of border around points title for plot, defaults to cell_color parameter title show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols

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list of saving parameters from all_plots_save_function

```
background_color
```

color of plot background

```
axis_text
                  size of axis text
axis_title
                  size of axis title
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
                  cowplot param: relative width
cow_rel_w
cow_align
                  cowplot param: how to align
                  show plot
show_plot
return_plot
                  return ggplot object
save_plot
                  directly save the plot [boolean]
```

Details

save_param

Description of parameters, see dimPlot2D. For 3D plots see plotTSNE_3D

Value

ggplot

Examples

```
plotTSNE(gobject)
```

plotTSNE_2D

plotTSNE_2D

Description

Short wrapper for tSNE visualization

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

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Arguments

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

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point_shape point with border or not (border or no_border)

point_size size of point (cell)

point_border_col

color of border around points

point_border_stroke

stroke size of border around points

title title for plot, defaults to cell_color parameter

show_legend show legend

 ${\tt legend_symbol_size}$

size of legend symbols

background_color

color of plot background

axis_text size of axis text
axis_title size of axis title

cow_n_col cowplot param: how many columns
cow_rel_h cowplot param: relative height
cow_rel_w cowplot param: relative width
cow_align cowplot param: how to align

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

Details

Description of parameters, see dimPlot2D. For 3D plots see plotTSNE_3D

Value

ggplot

Examples

plotTSNE_2D(gobject)

plotTSNE_3D plotTSNE_3D

Description

Visualize cells according to dimension reduction coordinates

plotTSNE_3D 227

```
Usage
```

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

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

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show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

Details

Description of parameters.

Value

plotly

Examples

```
plotTSNE_3D(gobject)
```

plotUMAP plotUMAP

Description

Short wrapper for UMAP visualization

Usage

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

Arguments

gobject giotto object

dim_reduction_name

dimension reduction name

 ${\tt default_save_name}$

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

groub_by create multiple plots based on cell annotation column

group_by_subset

subset the group_by factor column

dim1_to_use dimension to use on x-axis dim2_to_use dimension to use on y-axis

spat_enr_names names of spatial enrichment results to include

show_NN_network

show underlying NN network

nn_network_to_use

type of NN network to use (kNN vs sNN)

cell_color color for cells (see details)

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color_as_factor convert color column to factor cell_color_code named vector with colors cell_color_gradient vector with 3 colors for numeric data gradient_midpoint midpoint for color gradient gradient_limits vector with lower and upper limits select_cell_groups select subset of cells/clusters based on cell_color parameter select_cells select subset of cells based on cell IDs show_other_cells display not selected cells other_cell_color color of not selected cells other_point_size size of not selected cells show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels edge_alpha column to use for alpha of the edges point_shape point with border or not (border or no_border) size of point (cell) point_size point_border_col color of border around points point_border_stroke stroke size of border around points title title for plot, defaults to cell_color parameter show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols background_color color of plot background axis_text size of axis text axis_title size of axis title cow_n_col cowplot param: how many columns cow_rel_h cowplot param: relative height cow_rel_w cowplot param: relative width

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```
cow_align cowplot param: how to align
```

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

Details

Description of parameters, see dimPlot2D. For 3D plots see plotUMAP_3D

Value

ggplot

Examples

```
plotUMAP(gobject)
```

plotUMAP_2D

plotUMAP_2D

Description

Short wrapper for UMAP visualization

Usage

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

```
gobject
                 giotto object
dim_reduction_name
                 dimension reduction name
default_save_name
                 default save name for saving, don't change, change save_name in save_param
groub_by
                 create multiple plots based on cell annotation column
group_by_subset
                 subset the group_by factor column
dim1_to_use
                 dimension to use on x-axis
dim2_to_use
                 dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
show_NN_network
                 show underlying NN network
```

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 ${\tt cell_color_gradient}$ vector with 3 colors for numeric data gradient_midpoint midpoint for color gradient gradient_limits vector with lower and upper limits select_cell_groups select subset of cells/clusters based on cell_color parameter select_cells select subset of cells based on cell IDs show_other_cells display not selected cells other_cell_color color of not selected cells other_point_size size of not selected cells show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points size of labels label_size label_fontface font of labels edge_alpha column to use for alpha of the edges point_shape point with border or not (border or no_border) point_size size of point (cell) point_border_col color of border around points point_border_stroke stroke size of border around points title for plot, defaults to cell_color parameter title show_legend show legend legend_text size of legend text legend_symbol_size size of legend symbols background_color color of plot background size of axis text axis_text

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

Details

Description of parameters, see dimPlot2D. For 3D plots see plotUMAP_3D

Value

ggplot

Examples

```
plotUMAP_2D(gobject)
```

plotUMAP_3D plotUMAP_3D

Description

Visualize cells according to dimension reduction coordinates

Usage

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

plotUMAP_3D 233

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

Details

Description of parameters.

Value

plotly

Examples

```
plotUMAP_3D(gobject)
```

Description

Visualize cells in network layer according to dimension reduction coordinates

Usage

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

Arguments

```
annotated\_network\_DT
```

annotated network data.table of selected cells

edge_alpha alpha of network edges

show_legend show legend gobject giotto object

Details

Description of parameters.

Value

ggplot

Examples

```
plot_network_layer_ggplot(gobject)
```

Description

Visualize cells in point layer according to dimension reduction coordinates

```
plot_point_layer_ggplot(
      ggobject,
      annotated_DT_selected,
      annotated_DT_other,
      cell_color = NULL,
      color_as_factor = T,
      cell_color_code = NULL,
      cell_color_gradient = c("blue", "white", "red"),
      gradient_midpoint = 0,
      gradient_limits = NULL,
      select_cell_groups = NULL,
      select_cells = NULL,
      point_size = 1,
      point_border_col = "black",
      point_border_stroke = 0.1,
      show_cluster_center = F,
      show_center_label = T,
      center_point_size = 4,
      center_point_border_col = "black",
      center_point_border_stroke = 0.1,
      label_size = 4,
      label_fontface = "bold",
      edge_alpha = NULL,
      show\_other\_cells = T,
      other_cell_color = "lightgrey",
      other_point_size = 0.5,
      show_legend = T
    )
Arguments
    annotated_DT_selected
                     annotated data.table of selected cells
    annotated_DT_other
                     annotated data.table of not selected cells
    cell_color
                     color for cells (see details)
    color_as_factor
                     convert color column to factor
    cell_color_code
                     named vector with colors
    cell_color_gradient
                     vector with 3 colors for numeric data
    gradient_midpoint
                     midpoint for color gradient
    gradient_limits
                     vector with lower and upper limits
    select_cell_groups
                     select subset of cells/clusters based on cell_color parameter
                     select subset of cells based on cell IDs
    select_cells
                     size of point (cell)
    point_size
```

```
point_border_col
                  color of border around points
point_border_stroke
                  stroke size of border around points
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
{\tt center\_point\_size}
                  size of center points
label_size
                  size of labels
label_fontface font of labels
edge_alpha
                  column to use for alpha of the edges
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  size of not selected cells
show_legend
                  show legend
                  giotto object
gobject
```

Details

Description of parameters.

Value

ggplot

Examples

```
plot_point_layer_ggplot(gobject)
```

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

Description

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

Usage

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

```
annotated_DT_selected
                  annotated data.table of selected cells
annotated_DT_other
                  annotated data.table of not selected cells
cell_color
                  color for cells (see details)
color_as_factor
                  convert color column to factor
cell_color_code
                  named vector with colors
cell_color_gradient
                  vector with 3 colors for numeric data
gradient_midpoint
                  midpoint for color gradient
gradient_limits
                  vector with lower and upper limits
select_cell_groups
                  select subset of cells/clusters based on cell_color parameter
select_cells
                  select subset of cells based on cell IDs
                  size of point (cell)
point_size
show_cluster_center
                  plot center of selected clusters
show_center_label
                  plot label of selected clusters
```

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

Details

Description of parameters.

Value

ggplot

Examples

```
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

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

Arguments

 $\operatorname{sdim} x$ x-axis dimension name (default = 'sdimx') sdimy y-axis dimension name (default = 'sdimy') cell_locations_metadata_selected annotated location from selected cells cell_locations_metadata_other annotated location from non-selected cells cell_color color for cells (see details) color_as_factor convert color column to factor cell_color_code named vector with colors cell_color_gradient vector with 3 colors for numeric data gradient_midpoint midpoint for color gradient gradient_limits vector with lower and upper limits select_cell_groups select subset of cells/clusters based on cell_color parameter select subset of cells based on cell IDs select_cells size of point (cell) point_size point_border_col color of border around points point_border_stroke stroke size of border around points show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels show_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,
  {\tt cell\_locations\_metadata\_selected},
  cell_locations_metadata_other,
  cell_color = NULL,
  color_as_factor = T,
  cell_color_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
  gradient_midpoint = NULL,
  gradient_limits = NULL,
  select_cell_groups = NULL,
  select_cells = NULL,
  point_size = 2,
  show_cluster_center = F,
  show_center_label = T,
  center_point_size = 4,
  label_size = 4,
  label_fontface = "bold",
  show_other_cells = T,
  other_cell_color = "lightgrey",
  other_point_size = 1,
  show_legend = TRUE
)
```

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

Details

Description of parameters.

Value

ggplot

Examples

```
plot_spat_point_layer_ggplot_noFILL(gobject)
```

Description

creat ggplot point layer for spatial coordinates without borders

Usage

```
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,
  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,
  show_legend = TRUE
```

```
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)
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
background_color
                  background color
vor_border_color
                  borde colorr of voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
show_legend
                  show legend
gobject
                  giotto object
```

Details

Description of parameters.

Value

ggplot

Examples

```
plot_spat_voronoi_layer_ggplot(gobject)
```

print.giotto 245

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
```

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.

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

Arguments

```
gobject Giotto object

sign_matrix Matrix of signature genes for each cell type / process

expression_values

expression values to use

reverse_log_scale

reverse expression values from log scale

logbase log base to use if reverse_log_scale = TRUE

output_enrichment

how to return enrichment output
```

Details

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

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

Value

data.table with enrichment results

See Also

```
make Sign Matrix Rank
```

Examples

```
rankEnrich(gobject)
```

```
rankSpatialCorGroups rankSpatialCorGroups
```

Description

Rank spatial correlated clusters according to correlation structure

```
rankSpatialCorGroups(
  gobject,
  spatCorObject,
  use_clus_name = NULL,
  show_plot = NA,
  return_plot = FALSE,
  save_plot = NA,
```

rank_binarize 247

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

Arguments

gobject giotto object

spatCorObject spatial correlation object

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

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

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)
```

Arguments

path

path to the expression matrix

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Details

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

Value

sparse matrix

Examples

```
readExprMatrix()
```

```
readGiottoInstructions
```

read Giot to Instrunctions

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)
```

removeCellAnnotation 249

remove Cell Annotation remove Cell Annotation

Description

removes cell annotation of giotto object

Usage

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

Arguments

gobject giotto object

columns names of columns to remove

return_gobject boolean: return giotto object (default = TRUE)

Details

if return_gobject = FALSE, it will return the cell metadata

Value

giotto object

Examples

removeCellAnnotation(gobject)

removeGeneAnnotation removeGeneAnnotation

Description

removes gene annotation of giotto object

Usage

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

Arguments

gobject giotto object

columns names of columns to remove

return_gobject boolean: return giotto object (default = TRUE)

Details

if return_gobject = FALSE, it will return the gene metadata

250 reshape_to_data_point

Value

giotto object

Examples

removeGeneAnnotation(gobject)

replaceGiottoInstructions

replace Giot to Instructions

Description

Function to replace all instructions from giotto object

Usage

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

Arguments

gobject giotto object

instructions new instructions (e.g. result from createGiottoInstructions)

Value

named vector with giotto instructions

Examples

replaceGiottoInstructions()

```
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)
```

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)
```

rowSums_giotto

rowSums_giotto

Description

 $rowSums_giotto$

Usage

rowSums_giotto(mymatrix)

252 runPCA

runPCA

runPCA

Description

runs a Principal Component Analysis

Usage

```
runPCA(
  gobject,
  expression_values = c("normalized", "scaled", "custom"),
  reduction = c("cells", "genes"),
  name = "pca",
  genes_to_use = NULL,
  return_gobject = TRUE,
  center = F,
  scale_unit = F,
  ncp = 100,
  ...
)
```

Arguments

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

reduction cells or genes

name arbitrary name for PCA run genes_to_use subset of genes to use for PCA

return_gobject boolean: return giotto object (default = TRUE)

center center data first (default = FALSE)

scale_unit scale features before PCA

ncp number of principal components to calculate
... additional parameters for PCA (see details)

Details

See PCA for more information about other parameters.

Value

giotto object with updated PCA dimension recuction

Examples

```
runPCA(gobject)
```

runtSNE 253

runtSNE runtSNE

Description

run tSNE

Usage

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

Arguments

```
gobject
                 giotto object
expression_values
                 expression values to use
reduction
                 cells or genes
dim_reduction_to_use
                 use another dimension reduction set as input
dim_reduction_name
                 name of dimension reduction set to use
dimensions_to_use
                 number of dimensions to use as input
                 arbitrary name for tSNE run
name
                 if dim_reduction_to_use = NULL, which genes to use
genes_to_use
return_gobject boolean: return giotto object (default = TRUE)
                 tSNE param: number of dimensions to return
dims
perplexity
                 tSNE param: perplexity
theta
                 tSNE param: theta
                 tSNE param: do PCA before tSNE (default = FALSE)
do_PCA_first
                 use of seed
set_seed
seed_number
                 seed number to use
                 additional tSNE parameters
```

254 runUMAP

Details

See Rtsne for more information about these and other parameters.

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

Value

giotto object with updated tSNE dimension recuction

Examples

```
runtSNE(gobject)
```

runUMAP

runUMAP

Description

run UMAP

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

runUMAP 255

Arguments

gobject giotto object

expression_values

expression values to use

reduction cells or genes

dim_reduction_to_use

use another dimension reduction set as input

dim_reduction_name

name of dimension reduction set to use

dimensions_to_use

number of dimensions to use as input

name arbitrary name for UMAP run

genes_to_use if dim_reduction_to_use = NULL, which genes to use

return_gobject boolean: return giotto object (default = TRUE)

n_neighborsn_componentsUMAP param: number of neighborsUMAP param: number of components

n_epochs UMAP param: number of epochs min_dist UMAP param: minimum distance

spread UMAP param: spread

set_seed use of seed

seed_number seed number to use

... additional UMAP parameters

Details

See umap for more information about these and other parameters.

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

Value

giotto object with updated UMAP dimension recuction

Examples

runUMAP(gobject)

256 screePlot

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"),
  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"
)
```

Arguments

gobject

name name of PCA object if available expression_values expression values to use reduction cells or genes genes_to_use subset of genes to use for PCA

giotto object

center center data before PCA

scale_unit scale features before PCA
ncp number of principal components to calculate

ylim y-axis limits on scree plot

verbose verobsity 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

selectPatternGenes 257

Details

Screeplot works by plotting the explained variance of each individual PC in a barplot allowing you to identify which PC does not show a significant contribution anymore (= 'elbow method').

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_pos_genes Top positively correlated genes.
top_neg_genes Top negatively correlated genes.
min_pos_cor Minimum positive correlation score to include a gene.
min_neg_cor Minimum negative correlation score to include a gene.
```

Details

Description.

Value

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

Examples

```
selectPatternGenes(gobject)
```

Description

helper function to select expression values

Usage

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

Arguments

gobject giotto object

values expression values to extract

Value

expression matrix

```
select\_spatialNetwork
```

Description

function to select a spatial network

Usage

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

Description

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

show,giotto-method 259

show, giotto-method show method for giotto class

Description

show method for giotto class

Usage

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

 $\verb|showClusterDendrogram| showClusterDendrogram|$

Description

Creates dendrogram for selected clusters.

Usage

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

Arguments

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

260 showClusterHeatmap

```
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 from all_plots_save_function

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

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

showGiottoInstructions 261

Arguments

gobject giotto object

expression_values

expression values to use

genes vector of genes to use, default to 'all' cluster_column name of column to use for clusters cor correlation score to calculate 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 from all_plots_save_function

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)

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

 $\verb|showGiottoInstructions()|\\$

262 showPattern

showPattern showPattern

Description

show patterns for 2D spatial data

Usage

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

Arguments

gobject giotto object

spatPatObj Output from detectSpatialPatterns

dimension dimension to plot

trim Trim ends of the PC values.

background_color

background color for plot

grid_border_color

color for grid

show_legend show legend of ggplot

show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

Value

ggplot

See Also

showPattern2D

Examples

```
showPattern(gobject)
```

showPattern2D 263

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)
```

264 showPattern3D

showPattern3D

showPattern3D

Description

show patterns for 3D spatial data

Usage

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

Arguments

```
gobject
                  giotto object
spatPatObj
                  Output from detectSpatialPatterns
dimension
                  dimension to plot
                  Trim ends of the PC values.
trim
background_color
                  background color for plot
grid_border_color
                  color for grid
                  show legend of plot
show_legend
point_size
                  adjust the point size
axis_scale
                  scale the axis
                  cutomize the scale of the axis
custom_ratio
x_ticks
                  the tick number of x_axis
                  the tick number of y_axis
y_ticks
z_ticks
                  the tick number of z_axis
```

showPatternGenes 265

```
show_plot show plot

return_plot return plot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

Value

plotly

Examples

```
showPattern3D(gobject)
```

showPatternGenes

showPatternGenes

Description

show genes correlated with spatial patterns

Usage

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

Arguments

```
gobject giotto object

spatPatObj Output from detectSpatialPatterns

dimension dimension to plot genes for.

top_pos_genes Top positively correlated genes.

top_neg_genes Top negatively correlated genes.

point_size size of points

return_DT if TRUE, it will return the data.table used to generate the plots
```

show_plot show plot

266 showProcessingSteps

```
return_plot return ggplot object
```

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function()

default_save_name

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

Value

ggplot

Examples

showPatternGenes(gobject)

showProcessingSteps sh

show Processing Steps

Description

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

Usage

showProcessingSteps(gobject)

Arguments

gobject giotto object

Value

list of processing steps and names

Examples

showProcessingSteps(gobject)

showSpatialCorGenes 267

showSpatialCorGenes showSpatialCorGenes

Description

Shows and filters spatially correlated genes

Usage

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

Arguments

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

Value

data.table with filtered information

Examples

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

268 signPCA

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"),
 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"
```

Arguments

gobject giotto object name name of PCA object if available method to use to identify significant PCs method 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 number of principal components to calculate ncp scree_ylim y-axis limits on scree plot number of interations for jackstraw jack_iter jack_threshold p-value threshold to call a PC significant jack_ylim y-axis limits on jackstraw plot

silhouetteRank 269

Details

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

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

multiple PCA results can be stored by changing the name parameter

Value

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

Examples

```
signPCA(gobject)
```

silhouetteRank

silhouetteRank

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.

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

270 spatCellCellcom

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)

```
sort_combine_two_DT_columns
```

 $sort_combine_two_DT_columns$

Description

fast sorting and pasting of 2 character columns

Usage

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

Examples

```
sort_combine_two_DT_columns()
```

spatCellCellcom spatCellCellcom

Description

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

spatCellCellcom 271

Usage

```
spatCellCellcom(
 gobject,
  spatial_network_name = "Delaunay_network",
  cluster_column = "cell_types",
  random_iter = 1000,
  gene_set_1,
  gene_set_2,
  log2FC_addendum = 0.1,
 min_observations = 2,
 adjust_method = c("fdr", "bonferroni", "BH", "holm", "hochberg", "hommel", "BY",
    "none"),
  adjust_target = c("genes", "cells"),
 do_parallel = TRUE,
  cores = NA,
  verbose = c("a little", "a lot", "none")
)
```

Arguments

```
gobject
                  giotto object to use
spatial_network_name
                  spatial network to use for identifying interacting cells
cluster_column cluster column with cell type information
random_iter
                  number of iterations
gene_set_1
                  first specific gene set from gene pairs
gene_set_2
                  second specific gene set from gene pairs
log2FC_addendum
                  addendum to add when calculating log2FC
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
do_parallel
                  run calculations in parallel with mclapply
                  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)
```

272 spatCellPlot

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,
  point_shape = c("border", "no_border", "voronoi"),
 point_size = 3,
 point_border_col = "black",
 point_border_stroke = 0.1,
  show_cluster_center = F,
  show_center_label = F,
  center_point_size = 4,
  center_point_border_col = "black",
  center_point_border_stroke = 0.1,
  label_size = 4,
  label_fontface = "bold",
  show_network = F,
  spatial_network_name = "Delaunay_network",
 network_color = NULL,
 network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
  grid_color = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 other_cells_alpha = 0.1,
  coord_fix_ratio = NULL,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
 background_color = "white",
  vor_border_color = "white",
```

spatCellPlot 273

```
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"
    )
Arguments
    gobject
                      giotto object
    show_image
                      show a tissue background image
    gimage
                      a giotto image
                      name of a giotto image
    image_name
    sdimx
                      x-axis dimension name (default = 'sdimx')
                      y-axis dimension name (default = 'sdimy')
    sdimy
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                      numeric cell annotation columns
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
                      select subset of cells based on cell IDs
    select_cells
    point_shape
                      shape of points (border, no_border or voronoi)
    point_size
                      size of point (cell)
    point_border_col
                      color of border around points
    point_border_stroke
                      stroke size of border around points
    show_cluster_center
                      plot center of selected clusters
    show_center_label
                      plot label of selected clusters
    center_point_size
                      size of center points
    label_size
                      size of labels
```

label_fontface font of labels

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```
show_network
                  show underlying spatial network
spatial_network_name
                  name of spatial network to use
network_color
                 color of spatial network
network_alpha
                  alpha of spatial network
show_grid
                  show spatial grid
spatial_grid_name
                  name of spatial grid to use
                  color of spatial grid
grid_color
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
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
```

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,
  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_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 = "Delaunay_network",
 network_color = NULL,
 network_alpha = 1,
  show\_grid = F,
  spatial_grid_name = "spatial_grid",
  grid_color = NULL,
  show_other_cells = T,
 other_cell_color = "lightgrey",
 other_point_size = 1,
 other_cells_alpha = 0.1,
  coord_fix_ratio = NULL,
  show_legend = T,
  legend_text = 8,
  legend_symbol_size = 1,
 background_color = "white",
  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 = "spatCellPlot2D"
    )
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')
                      y-axis dimension name (default = 'sdimy')
    sdimy
    spat_enr_names names of spatial enrichment results to include
    cell_annotation_values
                      numeric cell annotation columns
    cell_color_gradient
                      vector with 3 colors for numeric data
    gradient_midpoint
                      midpoint for color gradient
    gradient_limits
                      vector with lower and upper limits
    select_cell_groups
                      select subset of cells/clusters based on cell_color parameter
                      select subset of cells based on cell IDs
    select_cells
    point_shape
                      shape of points (border, no_border or voronoi)
    point_size
                      size of point (cell)
    point_border_col
                      color of border around points
    point_border_stroke
                      stroke size of border around points
    show_cluster_center
                      plot center of selected clusters
    show_center_label
                      plot label of selected clusters
    center_point_size
                      size of center points
```

label_size

size of labels

label_fontface font of labels

```
show_network
                  show underlying spatial network
spatial_network_name
                  name of spatial network to use
network_color
                 color of spatial network
network_alpha
                  alpha of spatial network
show_grid
                  show spatial grid
spatial_grid_name
                  name of spatial grid to use
                  color of spatial grid
grid_color
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
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
```

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_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border", "voronoi"),
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
 dim_show_center_label = T,
 dim_center_point_size = 4,
 dim_center_point_border_col = "black",
 dim_center_point_border_stroke = 0.1,
 dim_label_size = 4,
  dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_center_point_border_col = "black",
  spat_center_point_border_stroke = 0.1,
  spat_label_size = 4,
  spat_label_fontface = "bold",
```

```
show_NN_network = F,
nn_network_to_use = "sNN",
nn_network_name = "sNN.pca",
dim_edge_alpha = 0.5,
spat_show_network = F,
spatial_network_name = "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,
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
                 a giotto image
gimage
                 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
```

dim1_to_use dimension to use on x-axis dim2_to_use dimension to use on y-axis sdimx = spatial dimension to use on x-axis sdimy = spatial dimension to use on y-axis cell_color_gradient vector with 3 colors for numeric data gradient_midpoint midpoint for color gradient gradient_limits vector with lower and upper limits select_cell_groups select subset of cells/clusters based on cell_color parameter select_cells select subset of cells based on cell IDs dim_point_shape spatial points with border or not (border or no_border) dim_point_size size of points in dim. reduction space dim_point_border_col border color of points in dim. reduction space ${\tt dim_point_border_stroke}$ 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_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 cow_rel_w cowplot param: relative width 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'
                  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]
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

```
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_border_col = "black",
dim_point_border_stroke = 0.1,
spat_point_shape = c("border", "no_border", "voronoi"),
spat_point_size = 1,
spat_point_border_col = "black",
spat_point_border_stroke = 0.1,
dim_show_cluster_center = F,
dim_show_center_label = T,
dim_center_point_size = 4,
dim_center_point_border_col = "black",
dim_center_point_border_stroke = 0.1,
dim_label_size = 4,
dim_label_fontface = "bold",
spat_show_cluster_center = F,
spat_show_center_label = F,
spat_center_point_size = 4,
spat_center_point_border_col = "black",
spat_center_point_border_stroke = 0.1,
spat_label_size = 4,
spat_label_fontface = "bold",
show_NN_network = F,
nn_network_to_use = "sNN",
nn_network_name = "sNN.pca",
dim_edge_alpha = 0.5,
spat\_show\_network = F,
spatial_network_name = "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,
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_border_col
                      border color of points in dim. reduction space
    dim_point_border_stroke
                      border stroke of points in dim. reduction space
```

shape of points (border, no_border or voronoi)

spat_point_shape

spat_point_size

size of spatial points

spat_point_border_col

border color of spatial points

spat_point_border_stroke

border stroke of spatial points

dim_show_cluster_center

show the center of each cluster

dim_show_center_label

provide a label for each cluster

dim_center_point_size

size of the center point

dim_center_point_border_col

border color of center point

dim_center_point_border_stroke

stroke size of center point

dim_label_size size of the center label

dim_label_fontface

font of the center label

spat_show_cluster_center

show the center of each cluster

spat_show_center_label

provide a label for each cluster

spat_center_point_size

size of the center point

spat_label_size

size of the center label

spat_label_fontface

font of the center label

show_NN_network

show underlying NN network

nn_network_to_use

type of NN network to use (kNN vs sNN)

nn_network_name

name of NN network to use, if show_NN_network = TRUE

dim_edge_alpha column to use for alpha of the edges

spat_show_network

show spatial network

spatial_network_name

name of spatial network to use

spat_network_color

color of spatial network

spat_show_grid show spatial grid

spatial_grid_name

name of spatial grid to use

spat_grid_color

color of spatial grid

show_other_cells

display not selected cells

```
other_cell_color
                  color of not selected cells
dim_other_point_size
                  size of not selected dim cells
spat_other_point_size
                  size of not selected spat cells
spat_other_cells_alpha
                  alpha of not selected spat cells
show_legend
                  show legend
legend_text
                  size of legend text
legend_symbol_size
                  size of legend symbols
dim_background_color
                  background color of points in dim. reduction space
spat_background_color
                  background color of spatial points
vor_border_color
                  border colorr for voronoi plot
vor_max_radius maximum radius for voronoi 'cells'
axis_text
                  size of axis text
                  size of axis title
axis_title
coord_fix_ratio
                  ratio for coordinates
                  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
                  directly save the plot [boolean]
save_plot
                  list of saving parameters from all_plots_save_function
save_param
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

Details

Description of parameters.

Value

ggplot

Examples

spatDimCellPlot2D(gobject)

spatDimGenePlot 287

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_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_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,
  axis_text = 8,
  axis_title = 8,
```

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```
cow_n_col = 2,
      cow_rel_h = 1,
      cow_rel_w = 1,
      cow_align = "h",
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatDimGenePlot"
    )
Arguments
    gobject
                     giotto object
    show_image
                     show a tissue background image
                     a giotto image
    gimage
                     name of a giotto image
    image_name
    expression_values
                     gene expression values to use
    plot_alignment direction to align plot
                     genes to show
    genes
    dim_reduction_to_use
                     dimension reduction to use
    dim_reduction_name
                     dimension reduction name
    dim1_to_use
                     dimension to use on x-axis
    dim2_to_use
                     dimension to use on y-axis
    dim_point_shape
                     dim reduction points with border or not (border or no_border)
    dim_point_size dim reduction plot: point size
    dim_point_border_col
                     color of border around points
    dim_point_border_stroke
                     stroke size of border around points
    show_NN_network
                     show underlying NN network
    nn_network_to_use
                     type of NN network to use (kNN vs sNN)
    network_name
                     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
    sdimx
                     spatial x-axis dimension name (default = 'sdimx')
                     spatial y-axis dimension name (default = 'sdimy')
    sdimy
```

name of spatial network to use

spatial_network_name

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```
spatial_grid_name
                 name of spatial grid to use
spat_point_shape
                 spatial points with border or not (border or no_border)
spat_point_size
                 spatial plot: point size
spat_point_border_col
                 color of border around points
spat_point_border_stroke
                 stroke size of border around points
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'
axis_text
                 size of axis text
axis_title
                 size of axis title
cow_n_col
                 cowplot param: how many columns
cow_rel_h
                 cowplot param: relative height
                 cowplot param: relative width
cow_rel_w
cow_align
                 cowplot param: how to align
show_plot
                 show 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
```

Details

Description of parameters.

Value

ggplot

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See Also

```
spatDimGenePlot3D
```

Examples

```
spatDimGenePlot(gobject)
```

spatDimGenePlot2D

spatDimGenePlot2D

Description

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

```
spatDimGenePlot2D(
 gobject,
 show_image = F,
 gimage = NULL,
  image_name = "image",
 expression_values = c("normalized", "scaled", "custom"),
 plot_alignment = c("vertical", "horizontal"),
 dim_reduction_to_use = "umap",
 dim_reduction_name = "umap",
 dim1_to_use = 1,
 dim2\_to\_use = 2,
 dim_point_shape = c("border", "no_border"),
 dim_point_size = 1,
 dim_point_border_col = "black",
  dim_point_border_stroke = 0.1,
  show_NN_network = F,
  show_spatial_network = F,
  show_spatial_grid = F,
 nn_network_to_use = "sNN",
 network_name = "sNN.pca",
 edge_alpha_dim = NULL,
  scale_alpha_with_expression = FALSE,
  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_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,
```

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```
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,
      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
                     dimension to use on x-axis
   dim1_to_use
    dim2_to_use
                     dimension to use on y-axis
    dim_point_shape
                     dim reduction points with border or not (border or no_border)
    dim_point_size dim reduction plot: point size
   dim_point_border_col
                     color of border around points
    dim_point_border_stroke
                     stroke size of border around points
    show_NN_network
                     show underlying NN network
   nn_network_to_use
                     type of NN network to use (kNN vs sNN)
                     name of NN network to use, if show_NN_network = TRUE
    network_name
```

edge_alpha_dim dim reduction plot: column to use for alpha of the edges

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```
scale_alpha_with_expression
                  scale expression with ggplot alpha parameter
                  spatial x-axis dimension name (default = 'sdimx')
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_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
cow_n_col
                  cowplot param: how many columns
                  cowplot param: relative height
cow_rel_h
                  cowplot param: relative width
cow_rel_w
                  cowplot param: how to align
cow_align
show_legend
                  show legend
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'
                  size of axis text
axis_text
                  size of axis title
axis_title
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
```

Details

Description of parameters.

spatDimGenePlot3D 293

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,
  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",
```

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```
spatial_network_alpha = 0.5,
      show_spatial_grid = F,
      spatial_grid_name = "spatial_grid",
      spatial_grid_color = NULL,
      spatial_grid_alpha = 0.5,
      spatial_point_size = 3,
      legend_text_size = 12,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
      x_{ticks} = NULL,
     y_ticks = NULL,
      z_ticks = NULL,
      show_plot = NA,
     return_plot = NA,
      save_plot = NA,
      save_param = list(),
     default_save_name = "spatDimGenePlot3D"
   )
Arguments
                    giotto object
   gobject
   expression_values
                    gene expression values to use
   plot_alignment direction to align plot
   dim_reduction_to_use
                    dimension reduction to use
   dim_reduction_name
                    dimension reduction name
   dim1_to_use
                    dimension to use on x-axis
                    dimension to use on y-axis
   dim2_to_use
   dim3_to_use
                    dimension to use on z-axis
   genes
                    genes to show
   show_NN_network
                    show underlying NN network
   nn_network_to_use
                    type of NN network to use (kNN vs sNN)
                    name of NN network to use, if show_NN_network = TRUE
   network_name
   dim_point_size dim reduction plot: point size
    spatial_network_name
                    name of spatial network to use
   spatial_grid_name
                    name of spatial grid to use
   spatial_point_size
                    spatial plot: point size
                    show plots
   show_plot
                    return plotly object
   return_plot
    save_plot
                    directly save the plot [boolean]
```

```
save_param list of saving parameters from all_plots_save_function

default_save_name

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

edge_alpha_dim dim reduction plot: column to use for alpha of the edges

scale_alpha_with_expression

scale expression with ggplot alpha parameter

point_size size of point (cell)

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_border_col = "black",
dim_point_border_stroke = 0.1,
spat_point_shape = c("border", "no_border", "voronoi"),
spat_point_size = 1,
spat_point_border_col = "black",
spat_point_border_stroke = 0.1,
dim_show_cluster_center = F,
dim_show_center_label = T,
dim_center_point_size = 4,
dim_center_point_border_col = "black",
dim_center_point_border_stroke = 0.1,
dim_label_size = 4,
dim_label_fontface = "bold",
spat_show_cluster_center = F,
spat_show_center_label = F,
spat_center_point_size = 4,
spat_label_size = 4,
spat_label_fontface = "bold",
show_NN_network = F,
nn_network_to_use = "sNN",
network_name = "sNN.pca",
nn_network_alpha = 0.05,
show_spatial_network = F,
spat_network_name = "spatial_network",
spat_network_color = "blue",
spat_network_alpha = 0.5,
show_spatial_grid = F,
spat_grid_name = "spatial_grid",
spat_grid_color = "blue",
show_other_cells = T,
other_cell_color = "lightgrey",
dim_other_point_size = 1,
spat_other_point_size = 1,
spat_other_cells_alpha = 0.5,
dim_show_legend = F,
spat_show_legend = F,
legend_text = 8,
legend_symbol_size = 1,
dim_background_color = "white",
spat_background_color = "white",
vor_border_color = "white",
vor_max_radius = 200,
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 a tissue background image show_image a giotto image gimage name of a giotto image image_name 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_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_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) network_name name of NN network to use, if show_NN_network = TRUE 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'
axis_text
                  size of axis text
axis_title
                  size of axis title
show_plot
                  show plot
return_plot
                  return ggplot object
                  directly save the plot [boolean]
save_plot
save_param
                  list of saving parameters from all_plots_save_function
default_save_name
                  default save name for saving, don't change, change save_name in save_param
```

Details

Description of parameters.

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_border_col = "black",
 dim_point_border_stroke = 0.1,
  spat_point_shape = c("border", "no_border", "voronoi"),
  spat_point_size = 1,
  spat_point_border_col = "black",
  spat_point_border_stroke = 0.1,
 dim_show_cluster_center = F,
 dim_show_center_label = T,
 dim_center_point_size = 4,
 dim_center_point_border_col = "black",
 dim_center_point_border_stroke = 0.1,
  dim_label_size = 4,
 dim_label_fontface = "bold",
  spat_show_cluster_center = F,
  spat_show_center_label = F,
  spat_center_point_size = 4,
  spat_label_size = 4,
  spat_label_fontface = "bold",
  show_NN_network = F,
```

```
nn_network_to_use = "sNN",
network_name = "sNN.pca",
nn_network_alpha = 0.05,
show_spatial_network = F,
spat_network_name = "spatial_network",
spat_network_color = "blue",
spat_network_alpha = 0.5,
show_spatial_grid = F,
spat_grid_name = "spatial_grid",
spat_grid_color = "blue",
show_other_cells = T,
other_cell_color = "lightgrey",
dim_other_point_size = 1,
spat_other_point_size = 1,
spat_other_cells_alpha = 0.5,
dim\_show\_legend = F,
spat\_show\_legend = F,
legend_text = 8,
legend_symbol_size = 1,
dim_background_color = "white",
spat_background_color = "white",
vor_border_color = "white",
vor_max_radius = 200,
axis_text = 8,
axis_title = 8,
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spatDimPlot2D"
```

Arguments

)

```
giotto object
gobject
show_image
                  show a tissue background image
                  a giotto image
gimage
                  name of a giotto image
image_name
plot_alignment direction to align plot
dim_reduction_to_use
                  dimension reduction to use
dim_reduction_name
                  dimension reduction name
dim1_to_use
                  dimension to use on x-axis
dim2_to_use
                  dimension to use on y-axis
sdimx
                  = spatial dimension to use on x-axis
sdimy
                  = spatial dimension to use on y-axis
spat_enr_names names of spatial enrichment results to include
cell_color
                  color for cells (see details)
```

color_as_factor convert color column to factor cell_color_code named vector with colors cell_color_gradient vector with 3 colors for numeric data gradient_midpoint midpoint for color gradient gradient_limits vector with lower and upper limits select_cell_groups select subset of cells/clusters based on cell color parameter select subset of cells based on cell IDs select_cells dim_point_shape point with border or not (border or no_border) dim_point_size size of points in dim. reduction space dim_point_border_col border color of points in dim. reduction space dim_point_border_stroke border stroke of points in dim. reduction space spat_point_shape shape of points (border, no_border or voronoi) spat_point_size size of spatial points spat_point_border_col border color of spatial points spat_point_border_stroke border stroke of spatial points dim_show_cluster_center show the center of each cluster dim_show_center_label provide a label for each cluster dim_center_point_size size of the center point dim_center_point_border_col border color of center point dim_center_point_border_stroke stroke size of center point dim_label_size size of the center label dim_label_fontface font of the center label spat_show_cluster_center show the center of each cluster spat_show_center_label provide a label for each cluster spat_center_point_size size of the center point spat_label_size

size of the center label

spat_label_fontface font of the center label show_NN_network show underlying NN network nn_network_to_use type of NN network to use (kNN vs sNN) name of NN network to use, if show_NN_network = TRUE network_name nn_network_alpha column to use for alpha of the edges $show_spatial_network$ show spatial network spat_network_name name of spatial network to use spat_network_color color of spatial network show_spatial_grid show spatial grid spat_grid_name name of spatial grid to use spat_grid_color color of spatial grid show_other_cells display not selected cells other_cell_color color of not selected cells dim_other_point_size size of not selected dim cells spat_other_point_size size of not selected spat cells spat_other_cells_alpha alpha of not selected spat cells dim_show_legend show legend of dimension reduction plot spat_show_legend show legend of spatial plot legend_text size of legend text legend_symbol_size size of legend symbols dim_background_color background color of points in dim. reduction space spat_background_color background color of spatial points vor_border_color border colorr for voronoi plot vor_max_radius maximum radius for voronoi 'cells' axis_text size of axis text axis_title size of axis title show_plot show plot

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
    dim3_to_use
                    dimension to use on z-axis
                    = spatial dimension to use on x-axis
    sdimx
    sdimy
                    = spatial dimension to use on y-axis
    sdimz
                    = spatial dimension to use on z-axis
    show_NN_network
                    show underlying NN network
    nn_network_to_use
                    type of NN network to use (kNN vs sNN)
                    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

spatGenePlot 307

Details

Description of parameters.

Value

plotly

Examples

```
spatDimPlot3D(gobject)
```

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"),
 cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 show_network = F,
 network_color = NULL,
  spatial_network_name = "Delaunay_network",
 edge_alpha = NULL,
 show\_grid = F,
 grid_color = NULL,
```

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```
spatial_grid_name = "spatial_grid",
midpoint = 0,
scale_alpha_with_expression = FALSE,
point_shape = c("border", "no_border", "voronoi"),
point_size = 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,
axis_text = 8,
axis_title = 8,
cow_n_col = 2,
cow_rel_h = 1,
cow_rel_w = 1,
cow_align = "h",
show_plot = NA,
return_plot = NA,
save_plot = NA,
save_param = list(),
default_save_name = "spatGenePlot"
```

Arguments

```
gobject
                  giotto object
show_image
                  show a tissue background image
gimage
                  a giotto image
                  name of a giotto image
image_name
sdimx
                  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
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 spatial grid
show_grid
grid_color
                  color of spatial grid
```

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

Details

Description of parameters.

Value

ggplot

See Also

```
spatGenePlot3D and spatGenePlot2D
```

Examples

```
spatGenePlot(gobject)
```

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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"),
 genes,
  cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
 show_network = F,
 network_color = NULL,
  spatial_network_name = "Delaunay_network",
  edge_alpha = NULL,
 show\_grid = F,
 grid_color = NULL,
  spatial_grid_name = "spatial_grid",
 midpoint = 0,
  scale_alpha_with_expression = FALSE,
 point_shape = c("border", "no_border", "voronoi"),
  point_size = 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,
 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"
```

spatGenePlot2D 311

Arguments

cow_rel_h

gobject giotto object show_image show a tissue background image gimage a giotto image image_name name of a giotto image x-axis dimension name (default = 'sdimx') sdimx y-axis dimension name (default = 'sdimy') sdimy expression_values gene expression values to use genes to show genes 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 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 midpoint expression midpoint scale_alpha_with_expression scale expression with ggplot alpha parameter shape of points (border, no_border or voronoi) point_shape point_size size of point (cell) point_border_col color of border around points point_border_stroke stroke size of border around points show_legend show legend 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' axis_text size of axis text axis_title size of axis title cowplot param: how many columns cow_n_col

cowplot param: relative height

312 spatGenePlot3D

```
cow_rel_w cowplot param: relative width

cow_align cowplot param: how to align

show_plot show plots

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

... parameters for cowplot::save_plot()
```

Details

Description of parameters.

Value

ggplot

See Also

```
spatGenePlot3D
```

Examples

```
spatGenePlot2D(gobject)
```

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,
```

spatGenePlot3D 313

```
genes_high_color = NULL,
      genes_mid_color = "white",
      genes_low_color = "blue",
      spatial_grid_name = "spatial_grid",
      point_size = 2,
      show_legend = T,
      axis_scale = c("cube", "real", "custom"),
      custom_ratio = NULL,
      x_{ticks} = NULL,
      y_ticks = NULL,
      z_ticks = NULL,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatGenePlot3D"
Arguments
    gobject
                     giotto object
    expression_values
                     gene expression values to use
    genes
                     genes to show
    show_network
                     show underlying spatial network
    network_color
                     color of spatial network
    spatial_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
                     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
                     color of spatial grid
    grid_color
                     expression midpoint
    midpoint
    scale_alpha_with_expression
                     scale expression with ggplot alpha parameter
                     parameters for cowplot::save_plot()
    . . .
```

314 spatialAEH

Details

Description of parameters.

Value

ggplot

Examples

```
spatGenePlot3D(gobject)
```

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 ...

spatialDE 315

Value

An updated giotto object

Examples

```
spatialAEH(gobject)
```

spatialDE

spatialDE

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
                  low/medium/high color scheme for plot
color
                  alpha value for significance
sig_alpha
unsig_alpha
                  alpha value for unsignificance
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]
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

316 Spatial_AEH

Details

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

Value

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

Examples

```
spatialDE(gobject)
```

Spatial_AEH

Spatial_AEH

Description

calculate automatic expression histology with spatialDE method

Usage

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

Arguments

gobject Giotto object
results output from spatial_DE
pattern_num the number of gene expression patterns
show_AEH show AEH plot
python_path specify specific path to python if required

Details

Description.

Spatial_DE 317

Value

a list or a dataframe of SVs

Examples

```
{\tt Spatial\_AEH(gobject)}
```

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)
```

318 spatNetwDistributions

```
{\tt spatNetwDistributions} \ \textit{spatNetwDistributionsDistance}
```

Description

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

Usage

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

Arguments

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

Examples

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

Value

```
ggplot plot
```

Examples

```
spatNetwDistributionsDistance(gobject)
```

```
spat Netw Distributions Kneighbors \\ spat Netw Distributions 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

```
Giotto object
gobject
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
                  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, alternatively change save_name in save_param
```

Value

ggplot plot

Examples

```
spatNetwDistributionsKneighbors(gobject)
```

spatPlot 321

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_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,
```

322 spatPlot

```
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 = "spatPlot"
    )
Arguments
   gobject
                     giotto object
                     show a tissue background image
    show_image
                     a giotto image
    gimage
    image_name
                     name of a giotto image
   group_by_subset
                     subset the group_by factor column
    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_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)
```

color of border around points

point_border_col

spatPlot 323

point_border_stroke stroke size of border around points show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels show_network show underlying spatial network spatial_network_name name of spatial network to use color of spatial network network_color network_alpha alpha of spatial network show_grid show spatial grid spatial_grid_name name of spatial grid to use color of spatial grid grid_color show_other_cells display not selected cells other_cell_color color of not selected cells other_point_size point size of not selected cells other_cells_alpha alpha of not selected cells coord_fix_ratio fix ratio between x and y-axis title of plot title 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' axis_text size of axis text axis_title size of axis title cow_n_col cowplot param: how many columns cow_rel_h cowplot param: relative height

cowplot param: relative width

cowplot param: how to align

cow_rel_w

cow_align

324 spatPlot2D

```
show_plot show plot

return_plot return ggplot object

save_plot directly save the plot [boolean]

save_param list of saving parameters from all_plots_save_function

default_save_name

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

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

Details

Description of parameters.

Value

ggplot

See Also

```
spatPlot3D
```

Examples

```
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_code = NULL,
  cell_color_gradient = c("blue", "white", "red"),
 gradient_midpoint = NULL,
 gradient_limits = NULL,
  select_cell_groups = NULL,
```

spatPlot2D 325

```
select_cells = NULL,
     point_shape = c("border", "no_border", "voronoi"),
     point_size = 3,
     point_border_col = "black",
     point_border_stroke = 0.1,
     show_cluster_center = F,
     show_center_label = F,
     center_point_size = 4,
     center_point_border_col = "black",
     center_point_border_stroke = 0.1,
     label_size = 4,
     label_fontface = "bold",
     show_network = F,
     spatial_network_name = 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,
     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 a tissue background image
   show_image
                   a giotto image
   gimage
```

name of a giotto image

subset the group_by factor column

image_name
group_by_subset

326 spatPlot2D

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_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_shape shape of points (border, no_border or voronoi) size of point (cell) point_size point_border_col color of border around points point_border_stroke stroke size of border around points show_cluster_center plot center of selected clusters show_center_label plot label of selected clusters center_point_size size of center points label_size size of labels label_fontface font of labels show_network show underlying spatial network spatial_network_name name of spatial network to use color of spatial network network_color 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

spatPlot2D 327

```
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'
axis_text
                  size of axis text
axis_title
                  size of axis title
cow_n_col
                  cowplot param: how many columns
cow_rel_h
                  cowplot param: relative height
cow_rel_w
                  cowplot param: relative width
cow_align
                  cowplot param: how to align
show_plot
                  show plot
return_plot
                  return ggplot object
                  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
```

create multiple plots based on cell annotation column

Details

Description of parameters.

Value

ggplot

groub_by

See Also

spatPlot3D

Examples

spatPlot2D(gobject)

328 spatPlot2D_single

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_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,
```

spatPlot2D_single 329

```
legend_symbol_size = 1,
      background_color = "white",
      vor_border_color = "white",
      vor_max_radius = 200,
      axis_text = 8,
      axis_title = 8,
      show_plot = NA,
      return_plot = NA,
      save_plot = NA,
      save_param = list(),
      default_save_name = "spatPlot2D_single"
    )
Arguments
                      giotto object
    gobject
                      show a tissue background image
    show_image
    gimage
                      a giotto image
    image_name
                      name of a giotto image
                      x-axis dimension name (default = 'sdimx')
    sdimx
    sdimy
                      y-axis dimension name (default = 'sdimy')
    spat_enr_names names of spatial enrichment results to include
    cell_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
                      size of point (cell)
    point_size
    point_border_col
                      color of border around points
    point_border_stroke
                      stroke size of border around points
    show_cluster_center
                      plot center of selected clusters
    show_center_label
                      plot label of selected clusters
```

center_point_size

size of center points

330 spatPlot2D_single

```
show_network
                  show underlying spatial network
spatial_network_name
                  name of spatial network to use
                  color of spatial network
network_color
network_alpha
                  alpha of spatial network
                  show spatial grid
show_grid
spatial_grid_name
                  name of spatial grid to use
grid_color
                  color of spatial grid
show_other_cells
                  display not selected cells
other_cell_color
                  color of not selected cells
other_point_size
                  point size of not selected cells
other_cells_alpha
                  alpha of not selected cells
coord_fix_ratio
                  fix ratio between x and y-axis
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'
axis_text
                  size of axis text
                  size of axis title
axis_title
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

Description of parameters.

Value

ggplot

label_size

size of labels

label_fontface font of labels

spatPlot3D 331

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,
 z_ticks = NULL,
 show_plot = NA,
 return_plot = NA,
  save_plot = NA,
 save_param = list(),
 default_save_name = "spat3D"
```

spatPlot3D

Arguments

gobject giotto object sdimx x-axis dimension name (default = 'sdimx') sdimy y-axis dimension name (default = 'sdimy') sdimz z-axis dimension name (default = 'sdimy') size of point (cell) point_size color for cells (see details) cell_color 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 spatial grid show_grid grid_color color of spatial grid spatial_grid_name name of spatial grid to use title title of plot show_legend show legend axis_scale the way to scale the axis custom_ratio customize the scale of the plot set the number of ticks on the x-axis x_ticks 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 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

spat_fish_func 333

Examples

```
spatPlot3D(gobject)
```

```
spat_fish_func
```

spat_fish_func

Description

performs fisher exact test

Usage

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

```
spat_OR_func
```

spat_OR_func

Description

calculate odds-ratio

Usage

```
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

```
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,
```

Arguments

```
gobject
                  giotto object to use
spatial_network_name
                  spatial network to use for identifying interacting cells
cluster_column cluster column with cell type information
random_iter
                  number of iterations
cell_type_1
                  first cell type
cell_type_2
                  second cell type
gene_set_1
                  first specific gene set from gene pairs
                  second specific gene set from gene pairs
gene_set_2
log2FC_addendum
                  addendum to add when calculating log2FC
min_observations
                  minimum number of interactions needed to be considered
                  which method to adjust p-values
adjust_method
adjust_target
                  adjust multiple hypotheses at the cell or gene level
verbose
                  verbose
```

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.

standardise_giotto 335

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)
```

standardise_giotto

standardise_giotto

Description

standardises a matrix

Usage

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

Arguments

x matrixcenter center datascale scale data

Value

standardized matrix

 ${\it stitch} Field Coordinates$

stitchFieldCoordinates

Description

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

336 stitchFieldCoordinates

Usage

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

Arguments

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

Details

Stitching of fields:

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

Value

Updated location dataframe with new X ['X_final'] and Y ['Y_final'] coordinates

```
stitchFieldCoordinates(gobject)
```

stitchTileCoordinates 337

stitchTileCoordinates

Description

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

Usage

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

Arguments

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

Details

•••

Examples

```
stitchTileCoordinates(gobject)
```

subClusterCells

subClusterCells

Description

subcluster cells

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

338 subClusterCells

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

Arguments

gobject

name name for new clustering result cluster_method clustering method to use

cluster_column cluster column to subcluster

selected_clusters

only do subclustering on these clusters

hvg_param parameters for calculateHVG

giotto object

hvg_min_perc_cells

threshold for detection in min percentage of cells

hvg_mean_expr_det

threshold for mean expression level in cells with detection

use_all_genes_as_hvg

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

min_nr_of_hvg minimum number of HVG, or all genes will be used as input for PCA

pca_param parameters for runPCA

nn_param parameters for parameters for createNearestNetwork

k_neighbors number of k for createNearestNetwork

resolution resolution gamma gamma omega omega

python_path specify specific path to python if required

nn_network_to_use

type of NN network to use (kNN vs sNN)

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

subsetGiotto 339

Value

giotto object with new subclusters appended to cell metadata

See Also

 ${\tt doLouvainCluster_multinet}, {\tt doLouvainCluster_community} \ and \ @see also \ {\tt doLeidenCluster}$

Examples

```
subClusterCells(gobject)
```

subsetGiotto

subsetGiot to

Description

subsets Giotto object including previous analyses.

Usage

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

Arguments

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

Value

giotto object

Examples

subsetGiotto(gobject)

340 subsetGiottoLocs

subsetGiottoLocs

subsetGiottoLocs

Description

subsets Giotto object based on spatial locations

Usage

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

Arguments

```
gobject giotto object

x_max maximum x-coordinate

x_min minimum x-coordinate

y_max maximum y-coordinate

y_min minimum y-coordinate

z_max maximum z-coordinate

z_min minimum z-coordinate

return_gobject return Giotto object
```

Details

if return_gobject = FALSE, then a filtered combined metadata data.table will be returned

Value

```
giotto object
```

```
subsetGiottoLocs(gobject)
```

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

Usage

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

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

342 viewHMRFresults

Value

data.frame with trendsceek spatial genes results

Examples

```
trendSceek(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
```

HMRF output from doHMRF k number of HMRF domains

betas_to_view results from different betas that you want to view

... paramters to visPlot()

Details

Description ...

Value

spatial plots with HMRF domains

See Also

```
visPlot
```

```
viewHMRFresults(gobject)
```

viewHMRFresults2D 343

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)
```

344 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 345

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
```

346 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 347

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
```

Arguments

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

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

Details

Description of parameters.

Value

ggplot

Examples

```
visDimGenePlot_2D_ggplot(gobject)
```

Description

Visualize cells and gene expression according to dimension reduction coordinates

Usage

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

Arguments

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

Details

Description of parameters.

visDimPlot 351

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,
```

352 visDimPlot

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

directory to save the plot

save_dir

visDimPlot_2D_ggplot 353

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 355

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

```
visDimPlot_2D_plotly(gobject)
```

```
visDimPlot_3D_plotly
```

Description

Visualize cells according to dimension reduction coordinates

Usage

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

Arguments

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

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

Details

Description of parameters.

Value

plotly

Examples

```
visDimPlot_3D_plotly(gobject)
```

visForceLayoutPlot visForceLayoutPlot

Description

Visualize cells according to forced layout algorithm coordinates

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

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

Arguments

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

360 visGenePlot

Details

Description of parameters.

Value

ggplot

Examples

visForceLayoutPlot(gobject)

visGenePlot

visGenePlot

Description

Visualize cells and gene expression according to spatial coordinates

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

visGenePlot 361

Arguments

gobject giotto object expression_values gene expression values to use genes to show genes genes_high_color color represents high gene expression genes_mid_color color represents middle gene expression genes_low_color color represents low gene expression show underlying spatial network show_network network_color color of spatial network spatial_network_name name of spatial network to use show spatial grid show_grid grid_color color of spatial grid spatial_grid_name name of spatial grid to use midpoint expression midpoint scale_alpha_with_expression scale expression with ggplot alpha parameter size of point (cell) point_size point_border_col color of border around points point_border_stroke stroke size of border around points show_legend show legend cow_n_col cowplot param: how many columns

cow_rel_h cowplot param: relative height cow_rel_w cowplot param: relative width cow_align cowplot param: how to align three mode to adjust axis scale axis_scale x_ticks number of ticks on x axis number of ticks on y axis y_ticks number of ticks on z axis z_ticks plot_method two methods of plot

show plots

Details

Description of parameters.

Value

ggplot or plotly

show_plots

Examples

```
visGenePlot(gobject)
```

```
visGenePlot_2D_ggplot visGenePlot_2D_ggplot
```

Description

Visualize cells and gene expression according to spatial coordinates

Usage

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

Arguments

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

visGenePlot_3D_plotly

```
genes_low_color
```

color represents low gene expression

show_network show underlying spatial network

network_color color of spatial network

spatial_network_name

name of spatial network to use

show_grid show spatial grid grid_color color of spatial grid

spatial_grid_name

name of spatial grid to use

midpoint expression midpoint

 ${\tt scale_alpha_with_expression}$

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

show_network network_color

show_grid

spatial_network_name

genes_high_color

genes_mid_color

genes_low_color

spatial_grid_name

point_size show_legend

axis_scale

x_ticks

y_ticks

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

show underlying spatial network

name of spatial network to use

color represents high gene expression

color represents middle gene expression

color represents low gene expression

name of spatial grid to use

three mode to adjust axis scale

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

size of point (cell)

show legend

color of spatial network

show spatial grid

visPlot 365

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,
```

366 visPlot

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

Arguments

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

visPlot_2D_ggplot 367

```
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,
```

368 visPlot_2D_ggplot

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

Arguments

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

visPlot_2D_ggplot 369

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

 ${\tt visPlot_2D_ggplot(gobject)}$

370 visPlot_2D_plotly

```
visPlot_2D_plotly
```

Description

Visualize cells according to spatial coordinates

Usage

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

Arguments

```
gobject giotto object

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

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

point_size size of point (cell)

cell_color color for cells (see details)

cell_color_code

named vector with colors

color_as_factor

convert color column to factor
```

visPlot_3D_plotly 371

```
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,
```

372 visPlot_3D_plotly

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

Arguments

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

visSpatDimGenePlot 373

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,
```

374 visSpatDimGenePlot

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

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edge_alpha_dim dim reduction plot: column to use for alpha of the edges scale_alpha_with_expression scale expression with ggplot alpha parameter

label_size size for the label

genes_low_color

color to represent low expression of gene

genes_high_color

color to represent high expression of gene

dim_point_size dim reduction plot: point size

spatial_network_name

name of spatial network to use

spatial_grid_name

name of spatial grid to use

spatial_point_size

spatial plot: point size

spatial_point_border_col

color of border around points

spatial_point_border_stroke

stroke size of border around points

legend_text_size

the size of the text in legend

axis_scale three modes to adjust axis scale ratio custom_ratio set the axis scale ratio on custom

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

midpoint size of point (cell)
point_size size of point (cell)

cow_n_col cowplot param: how many columns
cow_rel_h cowplot param: relative height
cow_rel_w cowplot param: relative width
cow_align cowplot param: how to align

show_legend show legend show_plot show plot

Details

Description of parameters.

Value

ggplot or plotly

Examples

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

visSpatDimGenePlot_2D visSpatDimGenePlot_2D

Description

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

Usage

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

Arguments

gobject giotto object

expression_values

gene expression values to use

plot_alignment direction to align plot

genes genes to show

dim_reduction_to_use

dimension reduction to use

dim_reduction_name

dimension reduction name

dim1_to_use dimension to use on x-axis dim2_to_use dimension to use on y-axis

point_size size of point (cell)

dim_point_border_col

color of border around points

dim_point_border_stroke

stroke size of border around points

show_NN_network

show underlying NN network

nn_network_to_use

type of NN network to use (kNN vs sNN)

 $network_name \qquad name \ of \ NN \ network \ to \ use, \ if \ show_NN_network = TRUE$

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

scale_alpha_with_expression

scale expression with ggplot alpha parameter

spatial_network_name

name of spatial network to use

spatial_grid_name

name of spatial grid to use

spatial_point_size

spatial plot: point size

spatial_point_border_col

color of border around points

 ${\tt spatial_point_border_stroke}$

stroke size of border around points

midpoint size of point (cell)

cow_n_col cowplot param: how many columns

cow_rel_h cowplot param: relative height cow_rel_w cowplot param: relative width cow_align cowplot param: how to align

show_legend show legend

dim_point_size dim reduction plot: point size

show_plot show plot

Details

Description of parameters.

Value

ggplot

Examples

```
visSpatDimGenePlot_2D(gobject)
```

```
visSpatDimGenePlot_3D visSpatDimGenePlot_3D
```

Description

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

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

y_ticks = NULL,

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

legend_text_size
toucher

spatial_network_alpha

show_spatial_grid

spatial_grid_name

spatial_grid_color

spatial_grid_alpha

text size of legend

color of spatial grid

alpha of spatial grid

alpha of spatial network

name of spatial grid to use

show spatial grid

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

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

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

Details

Description of parameters.

Value

ggplot

Examples

```
visSpatDimPlot_2D(gobject)
```

visSpatDimPlot_3D 385

visSpatDimPlot_3D

visSpatDimPlot_3D

Description

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

Usage

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

Arguments

gobject giotto object

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

Details

Description of parameters.

Value

plotly

writeHMRFresults 387

Examples

```
visSpatDimPlot_3D(gobject)
```

writeHMRFresults

writeHMRFresults

Description

write results from doHMRF to a data.table.

Usage

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

Arguments

gobject giotto object

HMRF output HMRF output from doHMRF

k k to write results for

betas_to_view results from different betas that you want to view

print_command see the python command

Value

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

Examples

```
writeHMRFresults(gobject)
```

Description

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

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

Arguments

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

Value

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

Description

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

Usage

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

Arguments

```
dim_reduction_cell

dimension reduction slot from giotto object

dim_red high level name of dimension reduction

dim_red_name specific name of dimension reduction to use

dim_red_rounding

numerical indicating how to round the coordinates

dim_red_rescale

numericals to rescale the coordinates

output_directory

directory where to save the files
```

Value

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

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

Description

write out numeric annotation data from a giotto object for the Viewer

Usage

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

Arguments

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

Value

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

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