# Package 'SPIAT'

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Type Package

Title Spatial Image Analysis of Tissues

Version 1.10.0

Description SPIAT (\*\*Sp\*\*atial \*\*I\*\*mage \*\*A\*\*nalysis of \*\*T\*\*issues) is an R package with a suite of data processing, quality control, visualization and data analysis tools. SPIAT is compatible with data generated from single-cell spatial proteomics platforms (e.g. OPAL, CODEX, MIBI, cellprofiler). SPIAT reads spatial data in the form of X and Y coordinates of cells, marker intensities and cell phenotypes. SPIAT includes six analysis modules that allow visualization, calculation of cell colocalization, categorization of the immune microenvironment relative to tumor areas, analysis of cellular neighborhoods, and the quantification of spatial heterogeneity, providing a comprehensive toolkit for spatial data analysis.

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## Description

SPIAT-package

SPIAT (\*\*Sp\*\*atial \*\*I\*\*mage \*\*A\*\*nalysis of \*\*T\*\*issues) is an R package with a suite of data processing, quality control, visualization and data analysis tools. SPIAT is compatible with data generated from single-cell spatial proteomics platforms (e.g. OPAL, CODEX, MIBI, cellprofiler). SPIAT reads spatial data in the form of X and Y coordinates of cells, marker intensities and cell phenotypes. SPIAT includes six analysis modules that allow visualization, calculation of cell colocalization, categorization of the immune microenvironment relative to tumor areas, analysis of cellular neighborhoods, and the quantification of spatial heterogeneity, providing a comprehensive toolkit for spatial data analysis.

SPIAT: Spatial Image Analysis of Tissues

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#### See Also

Useful links:

- https://trigosteam.github.io/SPIAT/
- Report bugs at https://github.com/trigosteam/SPIAT/issues

## Description

Calculate the difference of area under the curve (AUC) between two curves, normalised by the total area of the graph.

#### Usage

```
AUC_of_cross_function(df.cross)
```

## **Arguments**

df.cross

Data.frame. The output of calculate\_cross\_functions. Containing the positions of the two curves. Columns contain "r", "border" and "theo".

#### Value

A number

```
average_marker_intensity_within_radius 
 average_marker_intensity_within_radius
```

## **Description**

Calculates the average intensity of the target\_marker within a radius from the cells positive for the reference marker. Note that it pools all cells with the target marker that are within the specific radius of any reference cell. Results represent the average intensities within a radius, but not a vector of metrics for each cell.

## Usage

```
average_marker_intensity_within_radius(
   spe_object,
   reference_marker,
   target_marker,
   radius = 20
)
```

#### **Arguments**

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

reference_marker
String specifying the marker that is used for reference cells.

target_marker String specifying the marker to calculate its average intensity.

radius Numeric specifying the radius of search for cells around the reference cells.
```

#### Value

A single number is returned

## **Description**

Calculates the average minimum distance of all cells to their nearest cells in the input image.

## Usage

```
average_minimum_distance(spe_object)
```

## **Arguments**

spe\_object

SpatialExperiment object in the form of the output of format\_image\_to\_spe.

#### Value

A single number is returned

## **Examples**

```
average_minimum_distance(SPIAT::simulated_image)
```

```
average_nearest_neighbor_index
```

Average nearest neighbor index for point pattern (clustering or dispersion)

## **Description**

Calculate the the average nearest neighbor (ANN) index of a specified type of cells. The index indicates the clustering effect of a point pattern. The pattern can be clustering, random or dispersion.

```
average_nearest_neighbor_index(
  spe_object,
  reference_celltypes,
  feature_colname,
  p_val = 5e-06
)
```

#### **Arguments**

#### **Details**

ANN index is a statistical test to test for the presence of clusters of cells, (Clark and Evans, 1954). The ANN index evaluates the spatial aggregation or dispersion effect of objects based on the average distances between pairs of the nearest objects and can be used to test for the clustering of specific cell types (e.g. immune or tumor cells). Next, the z score and p-value of the ANN index is calculated to validate the significance of the pattern.

#### Value

A list with the ANN index, the pattern type and the corresponding p value

#### **Examples**

```
average_nearest_neighbor_index(SPIAT::defined_image, reference_celltypes =
"Tumour", feature_colname = "Cell.Type")
```

```
average_percentage_of_cells_within_radius

average_percentage_of_cells_within_radius
```

## Description

Calculates the average percentage of cells of a target cell type within a radius from the cells with a reference cell type. The calculation is done per reference cell, so runtime will depend on the number of reference cells present. Output is a single value (the mean for the image).

```
average_percentage_of_cells_within_radius(
   spe_object,
   reference_celltype,
   target_celltype,
   radius = 100,
   feature_colname
)
```

#### **Arguments**

#### Value

A numeric vector and a plot are returned

#### **Examples**

#### **Description**

Calculates the number and proportion of each cell type.

## Usage

```
calculate_cell_proportions(
  spe_object,
  reference_celltypes = NULL,
  celltypes_to_exclude = NULL,
  feature_colname = "Phenotype",
  plot.image = TRUE
)
```

#### Arguments

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe. reference\_celltypes

String Vector specifying reference cell types. If NULL (default), then the proportion of each cell type against all cells is returned. Alternatively, a custom vector of cell types can be used as input, and these will be used as the denominator in the calculation of the proportions.

```
celltypes_to_exclude
```

String Vector specifying cell types to exclude. For example "OTHER" will exclude that celltype from the Total. If NULL, all cell types are included.

feature\_colname

String. Column of cells to choose the cell type from (e.g. Phenotype, Cell.Type,

etc).

plot.image Boolean. Whether to plot the barplot of the cell percentages. By default is

TRUE.

#### Value

A data.frame is returned

## **Examples**

```
calculate_cell_proportions(SPIAT::defined_image, reference_celltypes = NULL,
celltypes_to_exclude = "Others", feature_colname="Cell.Type", plot.image = FALSE)
```

```
calculate_cross_functions
```

calculate\_cross\_functions

## Description

Compute and plot the cross functions between two specified cell types. This function implements the cross functions from [spatstat] package.

## Usage

```
calculate_cross_functions(
   spe_object,
   method = "Kcross",
   cell_types_of_interest,
   feature_colname,
   plot_results = TRUE,
   dist = NULL
)
```

#### **Arguments**

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

method String that is the method for dependence calculation. Options: "Gcross", "Kcross",

"Kcross.inhom", "Lcross", "Jcross". Default method is "Kcross".

cell_types_of_interest
```

String Vecto

String Vector. Cell types of interest.

feature\_colname

String that is the name of the column of the types.

plot\_results Boolean. TRUE if result to be plotted, FALSE if not. In either case, an object

with the results is returned

dist Number (OPTIONAL) The largest distance between two cell types at which K

function is evaluated. If NULL, use the default distances set by cross functions.

#### Value

An object of class "fv" defined in 'spatstat' package.

## **Examples**

```
df_cross <- calculate_cross_functions(SPIAT::defined_image,
method = "Kcross", cell_types_of_interest = c("Tumour","Immune3"),
feature_colname = "Cell.Type", dist = 100)</pre>
```

calculate\_distance\_to\_margin

calculate the distances of each cell to the margin

## Description

Returns a SPE object with the minimum distance from cells of interest (CoI) to the identified bordering cells.

#### Usage

```
calculate_distance_to_margin(spe_object)
```

#### **Arguments**

spe\_object

SpatialExperiment object. It should contain information of the detected bordering cells ('colData()' has 'Region' column).

## Value

An spe\_object with a 'Distance.To.Border' column is returned.

```
spe_border <- identify_bordering_cells(SPIAT::defined_image,
reference_cell = "Tumour", feature_colname = "Cell.Type", n_to_exclude = 10)
spe_dist <- calculate_distance_to_margin(spe_border)</pre>
```

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calculate\_entropy

calculate\_entropy

#### **Description**

If arg 'radius' is not specified, the function returns the entropy of the cell types of interest for the whole image. If arg 'radius' is specified, the function returns a data frame where each row is a reference cell and the columns stores the entropy of the cell types of interest in each circle of the reference cells.

## Usage

```
calculate_entropy(
  spe_object,
  cell_types_of_interest,
  feature_colname = "Phenotype",
  radius = NULL
)
```

#### **Arguments**

String Vector. Cell types of interest. If arg 'radius' is not NULL, the first cell type is considered as reference cell type. Circles of the specified radius will be drawn around the reference cells and the entropy of cell types will be calculated for each of the reference cells.

feature\_colname

String specifying the column the cell types are from.

radius

(OPTIONAL) Numeric. The maximum radius around a reference cell for another cell to be considered an interaction.

#### Value

A dataframe or a number depending on the argument radius

```
calculate_entropy(SPIAT::defined_image,
cell_types_of_interest = c("Immune1","Immune2"),
feature_colname = "Cell.Type")
```

```
calculate_minimum_distances_between_celltypes

calculate minimum distances between celltypes
```

#### **Description**

Returns the distance of the closest cell of a specific type from each reference cell.

## Usage

```
calculate_minimum_distances_between_celltypes(
  spe_object,
  feature_colname,
  cell_types_of_interest = NULL
)
```

#### **Arguments**

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

feature_colname

String of the feature column of cells to choose the cell types from (e.g. Cell.Type, Cell.Type2, etc).

cell_types_of_interest

String Vector of marker combinations to consider is FALSE.
```

#### Value

A data.frame is returned

## Examples

```
min_dists <- calculate_minimum_distances_between_celltypes(
SPIAT::defined_image, feature_colname = "Cell.Type",
cell_types_of_interest = c("Tumour","Immune1"))</pre>
```

#### **Description**

Returns the pairwise distances between cells of different types. If none of the cell types are found, it will print an error message and return a vector of NAs.

#### Usage

```
calculate_pairwise_distances_between_celltypes(
  spe_object,
  cell_types_of_interest = NULL,
  feature_colname
)
```

#### **Arguments**

String Vector containing cell types to be considered, if NULL, all cell type com-

binations will be calculated.

feature\_colname

String of the name the feature column with the cell types of interest to be considered.

#### Value

A data frame is returned.

#### **Examples**

```
calculate_pairwise_distances_between_celltypes(SPIAT::defined_image,
cell_types_of_interest = c("Tumour","Immune1"),
feature_colname = "Cell.Type")
```

## **Description**

Takes the result of <code>grid\_metrics</code> (a RasterLayer object) and calculates the percentage of the grid squares whose values are above or below a specified threshold.

## Usage

```
calculate_percentage_of_grids(raster_obj, threshold, above)
```

## **Arguments**

raster\_obj Raster object in the form of the output of grid\_metrics.

threshold Numeric. The threshold for defining the pattern.

above Boolean. Indicating whether the pattern is above (TRUE) or below (FALSE) the

threshold.

#### Value

A number is returned

#### **Examples**

```
grid <- grid_metrics(SPIAT::defined_image, FUN = calculate_entropy, n_split = 5,
cell_types_of_interest=c("Tumour","Immune3"), feature_colname = "Cell.Type")
calculate_percentage_of_grids(grid, threshold = 0.75, above = TRUE)</pre>
```

```
calculate\_proportions\_of\_cells\_in\_structure \\ calculate\_proportions\_of\_cells\_in\_structure
```

#### **Description**

Calculate the proportion of cells of interest in each defined tissue structure relative to all cells in each structure and relative to the same cell type in the whole image.

#### Usage

```
calculate_proportions_of_cells_in_structure(
  spe_object,
  cell_types_of_interest,
  feature_colname
)
```

#### Arguments

String. The name of the column where the cell types of interest are under.

## Value

A data.frame

```
spe_border <- identify_bordering_cells(SPIAT::defined_image,
reference_cell = "Tumour", feature_colname = "Cell.Type", n_to_exclude = 10)
spe_dist <- calculate_distance_to_margin(spe_border)
spe_structure <- define_structure(spe_dist,
cell_types_of_interest = c("Immune1","Immune2","Immune3"),
feature_colname = "Cell.Type", n_margin_layers = 5)
calculate_proportions_of_cells_in_structure(spe_structure,
cell_types_of_interest = c("Immune1","Immune3"),feature_colname="Cell.Type")</pre>
```

## Description

Takes the result of grid\_metrics (a RasterLayer object) and calculate its spatial autocorrelation.

#### Usage

```
calculate_spatial_autocorrelation(raster_obj, metric = "globalmoran", d = NULL)
```

## **Arguments**

raster\_obj Raster object in the form of the output of grid\_metrics.

metric String. The method for calculating spatial autocorrelation. Choose from "globalmoran" and "GearyC".

d Numeric. Upper bound local distance. The argument 'd2' from function moran. Default is NULL and the distance will be calculated automatically from the number of splits and the extent of the grid image.

#### Value

A number is returned

#### **Examples**

```
grid <- grid_metrics(SPIAT::defined_image, FUN = calculate_entropy,
n_split = 5, cell_types_of_interest=c("Tumour","Immune3"),
feature_colname = "Cell.Type")
calculate_spatial_autocorrelation(grid, metric = "globalmoran")</pre>
```

```
calculate\_summary\_distances\_between\_celltypes \\ calculate\_summary\_distances\_between\_celltypes
```

## Description

Returns the mean, median and standard deviation of the minimum/pairwise distances between phenotypes.

```
calculate_summary_distances_between_celltypes(df)
```

#### **Arguments**

df

Data.frame containing the distance output between cell types. The functions that generate the distances can be calculate\_minimum\_distances\_between\_celltypes and calculate\_pairwise\_distances\_between\_celltypes.

#### Value

A data frame is returned

#### **Examples**

## Description

Returns the mean, median and standard deviation of the distances between a specified cell type to the border.

#### Usage

```
calculate_summary_distances_of_cells_to_borders(
  spe_object,
  cell_types_of_interest,
  feature_colname = "Cell.Type"
)
```

## **Arguments**

feature\_colname

String specifying which column the interested cell types are from.

#### Value

A data frame is returned

#### **Examples**

```
spe_border <- identify_bordering_cells(SPIAT::defined_image,
reference_cell = "Tumour", feature_colname = "Cell.Type", n_to_exclude = 10)
spe_dist <- calculate_distance_to_margin(spe_border)
spe_structure <- define_structure(spe_dist, cell_types_of_interest =
c("Immune1","Immune2","Immune3"), feature_colname = "Cell.Type",
n_margin_layers = 5)
calculate_summary_distances_of_cells_to_borders(spe_structure,
cell_types_of_interest = c("Immune1","Immune3"),feature_colname = "Cell.Type")</pre>
```

```
composition\_of\_neighborhoods \\ composition\_of\_neighborhoods
```

#### **Description**

Returns a data.frame which contains the percentages of cells with a specific marker within each neighborhood. and the number of cells in the neighborhood.

#### **Usage**

```
composition_of_neighborhoods(spe_object, feature_colname)
```

## Arguments

```
spe_object SpatialExperiment that is the output of identify_neighborhoods. feature_colname

String. Column with cell types.
```

#### Value

A data.frame is returned

```
neighborhoods <- identify_neighborhoods(image_no_markers,
method = "hierarchical", min_neighborhood_size = 100,
cell_types_of_interest = c("Immune", "Immune1", "Immune2"), radius = 50,
feature_colname = "Cell.Type")
neighborhoods_vis <- composition_of_neighborhoods(neighborhoods,
feature_colname="Cell.Type")</pre>
```

18 crossing\_of\_crossK

#### **Description**

The function sweeps over circles of a range of radii surrounding reference cells and calculates the metrics at the radii. Metrics used with function need two conditions: 1) have a 'radius' parameter. 2) return a single number. For metrics that do not return a single number, users can wrap them in a new function that returns a number and then pass the new function to 'compute\_gradient()'.

#### Usage

```
compute_gradient(spe_object, radii, FUN, ...)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

radii Numeric Vector specifying the range of radii for the metrics to be calculated.

FUN Variable name specifying the metric.

... Arguments of FUN

#### Value

A list of the metrics under all radii

## **Examples**

```
gradient_positions <- c(30, 50, 100)
gradient_entropy <- compute_gradient(SPIAT::defined_image,
radii = gradient_positions, FUN = calculate_entropy,
cell_types_of_interest = c("Immune1","Immune2"),
feature_colname = "Cell.Type")</pre>
```

crossing\_of\_crossK

## **Description**

Determine if there is a crossing in the cross K curves, to further detect the existence of potential immune rings.

```
crossing_of_crossK(df.cross)
```

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## **Arguments**

df.cross

Data.frame. The output of calculate\_cross\_functions. Containing the positions of the two curves. Columns contain "r", "border" and "theo".

#### Value

A number. The percentage of the crossing position of the specified distance. Returns NA when the crossing happens too close to the y axis (<4

## **Examples**

defined\_image

SPE object of a simulated image with defined cell types based on marker combinations.

## Description

A dataset that contains a formatted spe object with cell ids, phenotypes, defined cell types in 'colData()' and marker intensities in 'assays()'. (The cell locations are the same with the cells in simulated\_image).

#### Usage

```
defined_image
```

## **Format**

An spe object. Assay contains 5 rows (markers) and 4951 columns (cells); colData contains 4951 rows (cells) and 3 columns (features).

#### See Also

```
simulated_image image_no_markers
```

20 define\_celltypes

define\_celltypes define\_celltypes

## **Description**

Define new cell types based on the existing cell types (categories) under a selected column (e.g. base on marker combinations under "Phenotype" column). This function will create a new column to store the new cell types.

## Usage

```
define_celltypes(
  spe_object,
  categories = NULL,
  category_colname = "Phenotype",
  names = NULL,
  new_colname = "Cell.Type",
  print_names = FALSE
)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe. Vector. Names of the old cell types to be defined; if NULL, the function will use categories predefined categories and names category\_colname

(Phenotype) String specifying the name of the column having the categories to be defined, by default "Phenotype".

names Vector of new names assigned to the selected categories; if NULL, the func-

tion will use predefined categories and names. Should be of the same length of

'categories'.

(Optional) String specifying the name of the column to be added, by default new\_colname

"Cell.Type".

print\_names (Optional) Boolean if the user wants the original and new names printed. Default

is FALSE.

#### Details

Users need to specify the names of the old cell categories and under which column the old cell categories exist. Then the users specify the names of the new cell types and the name of the new column to store the new cell types. Any cell categories that are not specified in 'categories' arg but present in the image will be defined as "Undefined" in the new column.

#### Value

An new SPE object is returned

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#### **Examples**

```
# the selected column is:
category_colname = "Phenotype"
# define the following marker combinations:
categories <- c("Tumour_marker", "Immune_marker1,Immune_marker2",
    "Immune_marker1,Immune_marker3",
    "Immune_marker1,Immune_marker2,Immune_marker4", "OTHER")
# the new defined cell names:
names = c("Tumour", "Immune1", "Immune2","Immune3", "Others")
# the new names are stored under this column:
new_colname <- "Cell.Type"

defined_spe <- define_celltypes(SPIAT::simulated_image,
    categories = categories, category_colname = category_colname, names = names,
    new_colname = new_colname)</pre>
```

define\_structure

define\_structure

## **Description**

After identifying the bordering cells of tissue regions and calculating the distances of each cell to the bordering cells, this function further identifies the cells that are located in the inside and outside of the identified regions, and in the internal and external margins. It also identifies particular types of cells that are infiltrated, stromal, internal margin or external margin cells.

## Usage

```
define_structure(
   spe_object,
   cell_types_of_interest,
   feature_colname = "Cell.Type",
   n_margin_layers = 5,
   margin_dist = NULL
)
```

#### **Arguments**

spe\_object

SpatialExperiment object that contains information of tumour bordering cells and cell distances to border ('colData()' has 'Region' and 'Distance.To.Border' columns).

 $cell\_types\_of\_interest$ 

String Vector of the names of the particular types of cells.

feature\_colname

String Specifying the column that contains the names of the immune cells.

n\_margin\_layers

Integer. The number of layers of cells that compose the internal/external margins. Default is 5.

margin\_dist Numeric. The width of the internal/external margins. Default is NULL. Only use when 'n\_margin\_layers' is NULL.

#### Value

A new spe object is returned. Under the 'Region' column, there will be potential categories including 'Border' - the bordering cells, 'Infiltrated.CoI' - cells of interest that present inside of the tissue regions, 'Inside' - cells within the regiona excluding the 'Infiltrated.CoI' cells and the cells at internal margin, 'Stromal.CoI' - cells of interest that present outside of the tissue regions, 'Outside' - cells outside of the tissue regions excluding the 'Stromal.CoI' cells, 'Internal.margin.CoI' - cells of interest that are in the internal margin of the tissue regions, 'Internal.margin' - cells in the internal margin of the tissue regions excluding the 'Internal.margin.CoI' cells, 'External.margin' - cells in the external margin of the tissue regions excluding the 'External.margin.CoI' cells.

#### **Examples**

```
spe_border <- identify_bordering_cells(SPIAT::defined_image,
reference_cell = "Tumour", feature_colname = "Cell.Type", n_to_exclude = 10)
spe_dist <- calculate_distance_to_margin(spe_border)
spe_structure <- define_structure(spe_dist,
cell_types_of_interest = c("Immune1","Immune2","Immune3"),
feature_colname = "Cell.Type", n_margin_layers = 5)
plot_cell_categories(spe_structure, feature_colname = "Structure")</pre>
```

dimensionality\_reduction\_plot

Dimensionality reduction plot

## **Description**

Generates the dimensionality reduction plots (UMAP or tSNE) based on marker intensities. Cells are grouped by the categories under the selected column.

```
dimensionality_reduction_plot(
   spe_object,
   plot_type = "UMAP",
   scale = TRUE,
   perplexity = 30,
   feature_colname
)
```

## Arguments

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

plot\_type String. Choose from "UMAP" and "TSNE".

scale Boolean. Whether scale the marker intensities.

perplexity Numeric. Perplexity parameter of the Rtsne function (should be positive and no

bigger than 3 \* perplexity < n - 1, where n is the number of cells).

feature\_colname

String. Specify the column name to group the cells.

#### Value

A plot

## **Examples**

```
dimensionality_reduction_plot(SPIAT::simulated_image, plot_type = "TSNE",
feature_colname = "Phenotype")
```

entropy\_gradient\_aggregated

The aggregated gradient of entropy and the peak of the gradient

#### **Description**

This function first calculates the entropy within circles of each reference cell at each radius. Then at each radius, the entropy of all circles surrounding each cell are aggregated into one number. The function sweeps over the specified radii and calculates the aggregated entropy under each radius.

#### Usage

```
entropy_gradient_aggregated(
   spe_object,
   cell_types_of_interest,
   feature_colname,
   radii
)
```

#### **Arguments**

String Vector. The cell types that the entropy is computed on.

feature\_colname

String. The column name of the interested cell types.

radii Numeric Vector. A vector of radii within a circle of a reference cell where the

entropy is computed on.

#### Value

A list of the gradient of entropy and the peak

#### **Examples**

```
gradient_pos <- seq(50, 500, 50)
gradient_results <- entropy_gradient_aggregated(SPIAT::defined_image,
cell_types_of_interest = c("Tumour","Immune3"),
feature_colname = "Cell.Type", radii = gradient_pos)
plot(1:10,gradient_results$gradient_df[1, 3:12])</pre>
```

```
format_cellprofiler_to_spe
```

Format a cellprofiler image into a SpatialExperiment object

## **Description**

Reads in spatial data in the form of cell coordinates, cell phenotypes (if available), and marker intensities and transforms to a SpatialExperiment object. The assay stores the intensity level of every marker (rows) for every cell (columns). Cell phenotype is stored under 'colData()'. Cell x and y coordinates are stored under 'spatialCoords()' Note that if the data does not include these parameters, we recommend adding it to the output from cellprofiler with NAs in columns.

## Usage

```
format_cellprofiler_to_spe(
  path = NULL,
  markers = NULL,
  intensity_columns_interest = NULL)
```

#### **Arguments**

path String of the path location cellprofiler csv file.

markers String Vector containing the markers used for staining.

intensity\_columns\_interest

String Vector with the names of the columns with the level of each marker. Column names must match the order of the 'markers' parameter.

#### **Details**

Note when specifying 'markers', please use "DAPI" to replace "DNA" due to implementation. The output data will include "DAPI" instead of "DNA".

#### Value

A SpatialExperiment object is returned

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#### **Examples**

```
path <- system.file("extdata", "tiny_cellprofiler.txt.gz", package = "SPIAT")
markers <- c("Marker1", "Marker2", "Marker3", "Marker4", "Marker5", "DAPI",
   "Marker6")
intensity_columns_interest <- c("Intensity_MeanIntensity_Marker1_rs",
   "Intensity_MeanIntensity_Marker2_rs", "Intensity_MeanIntensity_Marker3_rs",
   "Intensity_MeanIntensity_Marker4_rs", "Intensity_MeanIntensity_Marker5_rs",
   "Intensity_MeanIntensity_DAPI_rs", "Intensity_MeanIntensity_Marker6_rs")
formatted_cellprofiler <- format_cellprofiler_to_spe(path = path,
   markers = markers, intensity_columns_interest = intensity_columns_interest)</pre>
```

format\_codex\_to\_spe

Format a CODEX image into a SpatialExperiment object

#### Description

Reads in spatial data in the form of cell coordinates, cell phenotypes (if available), and marker intensities and transforms to a 'SpatialExperiment' object. The assay stores the intensity level of every marker (rows) for every cell (columns). Cell phenotype is stored under colData. Cell x and y coordinates are stored under 'spatialCoords()' field.

#### Usage

```
format_codex_to_spe(path = NULL, markers, path_to_codex_cell_phenotypes = NULL)
```

#### **Arguments**

```
path String of the path location of CODEX csv file.

markers String Vector containing the markers used for staining.

path_to_codex_cell_phenotypes

String of the path to the Cluster ID/Cell type file.
```

#### Value

A SpatialExperiment object is returned

```
path <- system.file("extdata", "tiny_codex.csv.gz", package = "SPIAT")
path_to_codex_cell_phenotypes <- system.file("extdata",
    "tiny_codex_phenotypes.txt.gz", package = "SPIAT")
markers <- c("CD45", "Ly6C", "CD27", "CD5", "CD79b")
formatted_codex <- format_codex_to_spe(path = path, markers = markers,
path_to_codex_cell_phenotypes = path_to_codex_cell_phenotypes)</pre>
```

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```
format_colData_to_spe
```

## **Description**

Format a data frame into a SpatialExperiment class where the count assay is empty every cell (columns), cell phenotypes are stored under colData() and cell coordinates are stored under spatial-Coords().

## Usage

```
format_colData_to_spe(df)
```

#### **Arguments**

df

Data frame that contains cell coordinates, phenotypes (if available) and other cell properties. The rownames should be cell ID

#### Value

An SpatialExperiment object

#### **Examples**

```
df <- data.frame(row.names = c("Cell_1", "Cell_2"), Cell.X.Position = c(2,5), Cell.Y.Position = c(3.3, 8), Phenotypes = c("CD3", "CD3, CD8")) spe <- format_colData_to_spe(df)
```

format\_halo\_to\_spe

Format a HALO image into a SpatialExperiment object

#### **Description**

Reads in HALO data in the form of cell coordinates, cell phenotypes (if available), and marker intensities and transforms to a 'SpatialExperiment' object. The assay stores the intensity level of every marker (rows) for every cell (columns). Cell x and y coordinates are stored under 'spatialCoords()'. Cell phenotype and other cell properties are stored under colData. The cell properties to be included are Cell.Area, Nucleus.Area and Cytoplasm.Area. Note that if the data does not include these parameters, we recommend adding it to the output from HALO with NAs in columns.

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#### Usage

```
format_halo_to_spe(
  path = NULL,
  markers = NULL,
  locations = NULL,
  dye_columns_interest = NULL,
  intensity_columns_interest = NULL)
```

#### **Arguments**

path String of the path location of HALO text file.

markers String Vector containing the markers used for staining.

locations (Optional) Vector containing the locations of markers used for staining. Loca-

tion can be either "Nucleus", "Cytoplasm" or "Membrane". This is used to select the Intensity column and can be used instead of 'intensity\_columns\_interest'.

dye\_columns\_interest

(Optional) Use if locations is not specified. Vector of names of the columns with the marker status (i.e. those indicating 1 or 0 for whether the cell is positive or negative for the marker). Column names must match the order of the 'markers' parameter.

intensity\_columns\_interest

(Optional) Use if locations is not specified. Vector with the names of the columns with the level of each marker. Column names must match the order of the 'markers' parameter.

#### Value

A SpatialExperiment object is returned

```
raw_halo_data <- system.file("extdata", "tiny_halo.csv.gz", package="SPIAT")
markers <- c("DAPI", "CD3", "PDL-1", "CD4", "CD8", "AMACR")
intensity_columns_interest <- c("Dye 1 Nucleus Intensity",
"Dye 2 Cytoplasm Intensity", "Dye 3 Membrane Intensity",
"Dye 4 Cytoplasm Intensity", "Dye 5 Cytoplasm Intensity",
"Dye 6 Cytoplasm Intensity")
dye_columns_interest <-c("Dye 1 Positive Nucleus", "Dye 2 Positive Cytoplasm",
"Dye 3 Positive Membrane", "Dye 4 Positive Cytoplasm",
"Dye 5 Positive Cytoplasm", "Dye 6 Positive Cytoplasm")
formatted_HALO <- format_halo_to_spe(path=raw_halo_data,markers=markers,
intensity_columns_interest=intensity_columns_interest,
dye_columns_interest=dye_columns_interest)</pre>
```

 $format_image_to_spe$ 

Format an image into a SpatialExperiment object

#### **Description**

Reads in spatial data in the form of cell coordinates, cell phenotypes (if available), and marker intensities and transforms to a SpatialExperiment object. The assay stores the intensity level of every marker (rows) for every cell (columns). Cell phenotype is stored under 'colData()'. Cell x and y coordinates are stored under 'spatialCoords()' field. The function can read in data in general format (manually constructed input), or data from other platforms including inForm, HALO, CODEX and cellprofiler. Alternatively, users can use the specific function for each format.

#### Usage

```
format_image_to_spe(
  format = "general",
  intensity_matrix = NULL,
  phenotypes = NULL,
  coord_x = NULL,
  coord_y = NULL,
  path = NULL,
  markers = NULL,
  locations = NULL,
  intensity_columns_interest = NULL,
  dye_columns_interest = NULL,
  path_to_codex_cell_phenotypes = NULL)
```

#### **Arguments**

format

String specifying the format of the data source. Default is "general" (REC-OMMENDED), where the cell phenotypes, coordinates and marker intensities are imported manually by the user. Other formats include "inForm", "HALO", "cellprofiler" and "CODEX".

intensity\_matrix

(Optional) For "general" format. A matrix of marker intensities or gene expression where the column names are the Cell IDs, and the rownames the marker.

phenotypes

(Optional) For "general" format. String Vector of cell phenotypes in the same order in which they appear in 'intensity\_matrix'. If no phenotypes available, then a vector of NAs can be used as input. Note that the combination of markers (e.g. CD3,CD4) needs to be used instead of the cell type name (e.g. helper T cells).

coord\_x

(Optional) For "general" format. Numeric Vector with the X coordinates of the cells. The cells must be in the same order as in the 'intensity\_matrix'.

coord\_y

(Optional) For "general" format. Numeric Vector with the Y coordinates of the cells. The cells must be in the same order as in the 'intensity\_matrix'.

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path (Optional) For formats other than "general". String of the path location of the

source file.

markers For formats other than "general". String Vector containing the markers used for

staining. These must be in the same order as the marker columns in the input file, and must match the marker names used in the input file. One of the markers

must be "DAPI".

locations (Optional) For "inForm" and "HALO". String Vector containing the locations

of markers used for staining. Location can be either "Nucleus", "Cytoplasm" or "Membrane". This is used to select the Intensity column and can be used instead

of 'intensity\_columns\_interest'.

intensity\_columns\_interest

(Optional) For "inForm" and "HALO", use if 'locations' is not specified. For "cellprofiler", mandatory. Vector with the names of the columns with the level of each marker. Column names must match the order of the 'markers' parameter.

dye\_columns\_interest

(Optional) For "HALO". Use if locations is not specified. Vector of names of the columns with the marker status (i.e. those indicating 1 or 0 for whether the cell is positive or negative for the marker). Column names must match the order of the 'markers' parameter.

path\_to\_codex\_cell\_phenotypes

(Optional) For "CODEX". String of the path to the Cluster ID/Cell type file.

#### **Details**

If the user inputs 'intensity\_matrix', please make sure the 'colnames' of the intensity matrix are the cell IDs. If the 'intensity\_matrix' is 'NULL', the function will automatically assign IDs to the cells.

Note for "cellprofiler" format, when specifying 'markers', please use "DAPI" to replace "DNA" due to implementation. The output data will include "DAPI" instead of "DNA".

The format of "Phenotype" column: For example, a cell positive for both "CD3" and "CD4" markers has the "CD3,CD4" \*\*cell phenotype\*\*. The phenotype has to be strictly formatted in such way where each positive marker has to be separated by a coma, with no space in between, and the order of the positive markers has to be the same as the order in the assay.

#### Value

A SpatialExperiment object is returned

#### See Also

```
format_inform_to_spe format_halo_to_spe format_codex_to_spe format_cellprofiler_to_spe
```

```
#Construct a marker intensity matrix (rows are markers, columns are cells)
intensity_matrix <- matrix(c(14.557, 0.169, 1.655, 0.054, 17.588, 0.229,
1.188, 2.074, 21.262, 4.206, 5.924, 0.021), nrow = 4, ncol = 3)
# define marker names as rownames
rownames(intensity_matrix) <- c("DAPI", "CD3", "CD4", "AMACR")</pre>
```

```
# define cell IDs as colnames
colnames(intensity_matrix) <- c("Cell_1", "Cell_2", "Cell_3")
# Construct a dummy metadata (phenotypes, x/y coordinates)
# the order of the elements in these vectors correspond to the cell order
# in `intensity matrix`
phenotypes <- c("OTHER", "AMACR", "CD3,CD4")
coord_x <- c(82, 171, 184)
coord_y <- c(30, 22, 38)

formatted_image <- format_image_to_spe(intensity_matrix=intensity_matrix,
phenotypes = phenotypes, coord_x = coord_x,coord_y = coord_y)</pre>
```

format\_inform\_to\_spe Format an inForm image into a SpatialExperiment object

## Description

Reads in inForm data in the form of cell coordinates, cell phenotypes (if available), and marker intensities and transforms to a SpatialExperiment object. The assay stores the intensity level of every marker (rows) for every cell (columns). Cell phenotype, x and y coordinates and other cell properties are stored under colData. The cell properties to include are Cell.Area, Nucleus.Area, Nucleus.Compactness, Nucleus.Axis.Ratio, and Cell.Axis.Ratio. Note that if the data does not include these parameters, we recommend adding it to the output from inForm with NAs in columns.

## Usage

```
format_inform_to_spe(
  path,
  markers,
  locations = NULL,
  intensity_columns_interest = NULL)
```

#### **Arguments**

path String of the path location of inForm text file.

markers String Vector containing the markers used for staining.

locations (Optional) String Vector containing the locations of markers used for staining.

Location can be either "Nucleus", "Cytoplasm" or "Membrane". This is used to select the Intensity column and can be used instead of 'intensity\_columns\_interest'.

 $intensity\_columns\_interest$ 

(Optional) Use if 'locations' is not specified. Vector with the names of the columns with the level of each marker. Column names must match the order

of the 'markers' parameter.

#### Value

A SpatialExperiment object is returned

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#### **Examples**

```
raw_inform_data<-system.file("extdata","tiny_inform.txt.gz",package="SPIAT")
markers <- c("DAPI", "CD3", "PD-L1", "CD4", "CD8", "AMACR")
locations <- c("Nucleus", "Cytoplasm", "Membrane", "Cytoplasm", "Cytoplasm")
formatted_inForm <- format_inform_to_spe(path=raw_inform_data,
markers=markers, locations=locations)</pre>
```

format\_spe\_to\_ppp

Format SPE object as a ppp object ('spatstat' package)

## Description

Formats an spe object into a ppp object which has the x,y coordinates, phenotypes as markers window specifies the range of x and y coordinates

#### Usage

```
format_spe_to_ppp(
   spe_object,
   window_pol = FALSE,
   feature_colname = "Phenotype"
)
```

## Arguments

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

window\_pol Optional Boolean Specifying if the window is polygon.

feature\_colname

String specifying the feature column of interest.

#### Value

A ppp object is returned (defined in 'spatstat' package)

```
ppp_object<-format_spe_to_ppp(SPIAT::defined_image,
feature_colname = "Cell.Type")
```

grid\_metrics

Split an image into grid and calculates a metric for each grid square

#### **Description**

Calculates a specified metric for each grid tile in the image and plots the metrics for the grid tiles.

## Usage

```
grid_metrics(spe_object, FUN, n_split, ...)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

FUN Variable name specifying the metric to be calculated.

n\_split Integer specifying the number of splits for the calculation of metrics. This number is the splits on each side (e.g. 'n\_split' = 3 means the image will be split into 9 tiles.)

... Arguments of FUN

#### Value

A list of the metrics of all grid tiles

## **Examples**

```
grid <- grid_metrics(SPIAT::defined_image, FUN = calculate_entropy, n_split = 5,
    cell_types_of_interest=c("Tumour","Immune3"), feature_colname = "Cell.Type")

identify_bordering_cells
    identify_bordering_cells</pre>
```

## **Description**

Identify the cells bordering a group of cells of a particular phenotype, and calculate the number of clustered groups of this cell type.

```
identify_bordering_cells(
   spe_object,
   reference_cell,
   feature_colname = "Cell.Type",
   ahull_alpha = NULL,
   n_to_exclude = 10,
   plot_final_border = TRUE
)
```

#### **Arguments**

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

reference_cell String. Cells of this cell type will be used for border detection.

feature_colname

String that specifies the column of 'reference_cell'.

ahull_alpha Number specifying the parameter for the alpha hull algorithm. The larger the number, the more cells will be included in one cell cluster.

n_to_exclude Integer. Clusters with cell count under this number will be deleted.

plot_final_border

Boolean if plot the identified bordering cells.
```

#### **Details**

The bordering cell detection algorithm is based on computing an alpha hull (Hemmer et al., 2020), a generalization of convex hull (Green and Silverman, 1979). The cells detected to be on the alpha hull are identified as the bordering cells.

#### Value

A new SPE object is returned. The SPE object has a 'Region' column with "Border", "Inside" and "Outside" categories. The returned object also has an attribute saving the number of clusters.

## **Examples**

```
spe_border <- identify_bordering_cells(SPIAT::defined_image,
  reference_cell = "Tumour", feature_colname = "Cell.Type", n_to_exclude = 10)
  n_clusters <- attr(spe_border, "n_of_clusters") # get the number of clusters
  n_clusters

identify_neighborhoods

identify_neighborhoods</pre>
```

#### **Description**

Uses Euclidean distances to identify neighborhoods of cells. Three clustering methods are available, including hierarchical clustering, dbscan, and (Rphenograph).

```
identify_neighborhoods(
   spe_object,
   method = "hierarchical",
   cell_types_of_interest,
   radius,
   min_neighborhood_size = 10,
```

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```
k = 100,
feature_colname,
no_pheno = NULL
)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

method String. The clustering method. Choose from "hierarchical", "dbscan" and "Rpheno-

graph". (Note Rphenograph function is not available for this version yet).

cell\_types\_of\_interest

String Vector of phenotypes to consider.

radius Numeric specifying the radius of search. Need to specify when 'method' is

"hierarchical" or "dbscan".

min\_neighborhood\_size

Numeric. The minimum number of cells within each cluster. Need to specify

when 'method' is "hierarchical" or "dbscan".

k Numeric. The parameter for "Rphenograph" method.

feature\_colname

String. Column from which the cell types are selected.

no\_pheno Cell type corresponding to cells without a known phenotype (e.g. "None",

"Other")

#### Value

An spe object and a plot is returned. The spe object contains information of the defined neighborhood under "Neighborhood" column. The cells of interest that do not form clusters are labelled "Free\_cell", cells not of interest are labelled 'NA'.

#### **Examples**

```
neighborhoods <- identify_neighborhoods(image_no_markers, method = "hierarchical",
min_neighborhood_size = 100, cell_types_of_interest = c("Immune", "Immune1", "Immune2"),
radius = 50, feature_colname = "Cell.Type")</pre>
```

image\_no\_markers

SPE object of a formatted image without marker intensities (simulated by 'spaSim' package)

## Description

A dataset that contains a formatted spe object with cell ids and cell types in 'colData()' and cell coordinates in 'spatialCoords()'. This dataset does not contain assays (marker intensities).

```
image_no_markers
```

image\_splitter 35

#### **Format**

An spe object. colData contains 4951 rows (cells) and 3 columns (features).

#### See Also

defined\_image simulated\_image

image\_splitter

Split a large image into sub images

## **Description**

Takes in an image in SpatialExperiment format, splits the image into specified sections and returns a list of SpatialExperiment objects. Users can choose to plot the cell positions in each sub image. Note that this function does not split the assay.

#### Usage

```
image_splitter(
   spe_object,
   number_of_splits,
   plot = FALSE,
   cut_labels = TRUE,
   colour_vector = NULL,
   minX = NULL,
   maxX = NULL,
   minY = NULL,
   maxY = NULL,
   feature_colname = "Phenotype"
)
```

#### **Arguments**

plot

spe\_object 'SpatialExperiment' object in the form of the output of format\_image\_to\_spe. number\_of\_splits

Numeric. specifying the number of segments (e.g. 2 = 2x2, 3 = 3x3).

Boolean. Specifies whether the splitted images should be printed in a pdf.

cut\_labels Boolean. Specifies whether to plot where the image had been segmented.

colour\_vector String Vector. If specified, the colours will be used for plotting. If NULL, colors

will be generated automatically.

minX Integer used to specify the minimum x boundary to be splitted.

maxX Integer used to specify the maximum x boundary to be splitted.

minY Integer used to specify the minimum y boundary to be splitted.

maxY Integer used to specify the maximum y boundary to be splitted.

feature\_colname

String specifying which column the colouring should be based on. Specify when 'plot' is TRUE. Default is "Phenotype".

#### Value

A list of spe objects is returned. Each data frame represents an image without assay data.

## **Examples**

```
split_image <- image_splitter(SPIAT::simulated_image, number_of_splits=3,
plot = FALSE)</pre>
```

## Description

Produces boxplots of marker levels for cells phenotyped as being positive for the marker, and those that where phenotyped as being negative.

## Usage

```
marker_intensity_boxplot(spe_object, marker, feature_colname = "Phenotype")
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

marker String. Marker being queried.

feature\_colname

String. Column containing marker information

#### Value

A plot is returned

```
marker_intensity_boxplot(SPIAT::simulated_image, "Immune_marker1")
```

## **Description**

Takes in the returned dataframe from marker\_threshold\_plot and generates a .pdf file containing scatter plots of actual intensity and predicted intensity for every marker.

#### Usage

```
marker_prediction_plot(predicted_data, marker)
```

## Arguments

```
predicted_data Output from predict_phenotypes.
marker String. Marker to plot
```

#### Value

A plot is returned

## **Examples**

```
predicted_result <- predict_phenotypes(spe_object = simulated_image, thresholds = NULL,
tumour_marker = "Tumour_marker",baseline_markers = c("Immune_marker1", "Immune_marker2",
"Immune_marker3", "Immune_marker4"), reference_phenotypes = TRUE)
marker_prediction_plot(predicted_result, marker = "Tumour_marker")
```

```
marker_surface_plot marker_surface_plot
```

## **Description**

Generates a 3D surface plot of the level of the selected marker. Note that the image is blurred based on the 'num\_splits' parameter.

```
marker_surface_plot(
   spe_object,
   num_splits,
   marker,
   x_position_min = NULL,
   x_position_max = NULL,
   y_position_min = NULL,
   y_position_max = NULL
)
```

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

num_splits Integer specifying the number of splits on the image, higher splits equal to higher resolution. Recommendation: 10-100

marker Marker to plot

x_position_min Integer specifying the minimum x boundary to be splitted

x_position_max Integer specifying the minimum y boundary to be splitted

y_position_max Integer specifying the maximum y boundary to be splitted
```

#### Value

A plot is returned

## **Examples**

```
marker_surface_plot(SPIAT::simulated_image, num_splits=15, marker="Immune_marker1")
```

```
marker_surface_plot_stack

marker_surface_plot_stack
```

## **Description**

Generates stacked 3D surface plots showing normalized intensity level of specified markers.

```
marker_surface_plot_stack(
   spe_object,
   num_splits,
   markers_to_plot,
   sep = 1,
   x_position_min = NULL,
   x_position_max = NULL,
   y_position_min = NULL,
   y_position_max = NULL
)
```

spe_object	SpatialExperiment object in the form of the output of format_image_to_spe.
num_splits	Integer specifying the number of splits on the image, higher splits equal to higher resolution. Recommendation: 10-100.
markers_to_plot	
	Vector of marker names for plotting.
sep	Integer specifying the distance separation between each surface plot. We recommend values in the 1-2 range.
x_position_min	Integer specifying the minimum x boundary to be splitted.
x_position_max	Integer specifying the maximum x boundary to be splitted.
y_position_min	Integer specifying the minimum y boundary to be splitted.
y_position_max	Integer specifying the maximum y boundary to be splitted.

#### Value

A plot is returned

## **Examples**

```
marker_surface_plot_stack(SPIAT::simulated_image, num_splits=15,
markers=c("Tumour_marker", "Immune_marker4"))
```

```
measure\_association\_to\_cell\_properties \\ measure\_association\_to\_cell\_properties
```

# Description

Plots the density or boxplot of a property of two cell celltypes or compares using t test/wilcoxon rank sum test.

```
measure_association_to_cell_properties(
   spe_object,
   property = "Cell.Area",
   celltypes,
   feature_colname = "Cell.Type",
   method = "density",
   Nucleus.Ratio = FALSE,
   log.scale = FALSE
)
```

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

property String that is the name of the column of interest.

celltypes String Vector of celltypes of interest.

feature\_colname

String that speficies the column of the cell types.

method String. The analysis to perform on the selected cell types and property. Options

are "density", "box", "t", "wilcox".

Nucleus.Ratio Boolean whether the ratio of the nucleus size is of interest.

log.scale Boolean whether to log the data.

#### Value

With method "box" or "density a plot is returned. With method "t" or "wilcox", the text output from the test are returned.

#### **Examples**

mixing\_score\_summary Calculate the (normalised) mixing score for interested cell types

## **Description**

Produces a data frame with mixing scores of input reference and target cells from a SpatialExperiment object. It calculates reference-target interactions and reference-reference interactions based on a radius. It derives the mixing score and the normalised mixing score. Function returns NA if the mixing score is being calculated between cells of the same type.

```
mixing_score_summary(
  spe_object,
  reference_celltype,
  target_celltype,
  radius = 20,
  feature_colname
)
```

feature\_colname

String specifying the column with the desired cell type annotations.

#### **Details**

The mixing score was originally defined as the number of immune-tumour interactions divided by the number of immune-immune interactions within a defined radius (Keren et al., 2018). The normalised mixing score normalises the immune-tumour interactions and immune-immune interactions within radius by the total number of immune-tumour and immune-immune interactions in the image, respectively. We have generalized this score to allow calculation of any two cell phenotypes defined by the user.

#### Value

A data frame of cell numbers, number of cell interactions, mixing scores, and normalised mixing scores. If there are no reference or target cells found in the image, or there are no reference cells found within the specified radius of any reference cells, the returned (normalised) mixing scores will be NA. If there are no target cells found within the radius of any reference cells, the returned (normalised) mixing scores will be 0.

#### **Examples**

#### **Description**

Calculates the number of cells of a target cell type within a pre-defined radius around cells of a reference cell type.

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#### Usage

```
number_of_cells_within_radius(
  spe_object,
  reference_celltype,
  target_celltype,
  radius = 20,
  feature_colname
)
```

## **Arguments**

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

reference_celltype
String. Cell type to be used for reference cells.

target_celltype
String. Cell type to be used for target cells.

radius Numeric. Radius around the reference cells.

feature_colname
```

String specifying the column with the desired cell type annotations.

#### Value

A list of dataframes with the number of target cells of each of the reference cells

#### **Examples**

## **Description**

Takes in a vector or radii and calculates the average intensity of a target marker using average\_intensity function. It plots the intensity level as a line graph.

```
plot_average_intensity(spe_object, reference_marker, target_marker, radii)
```

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## Arguments

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

reference_marker
String specifying the reference marker.

target_marker String specifying the marker to calculate its average intensity.

Numeric Vector specifying the search radius around reference cells.
```

#### Value

A plot is returned

## **Examples**

```
plot_average_intensity(SPIAT::simulated_image, reference_marker="Immune_marker3", target_marker="Immune_marker2", c(30, 35, 40, 45, 50, 75, 100))
```

```
plot_cell_categories plot_cell_categories
```

# Description

Produces a scatter plot of the cells of their x-y positions in the tissue. Cells are coloured categorically by phenotype. Cells not part of the phenotypes of interest will be coloured "lightgrey".

## Usage

```
plot_cell_categories(
   spe_object,
   categories_of_interest = NULL,
   colour_vector = NULL,
   feature_colname = "Cell.Type",
   cex = 1,
   layered = FALSE
)
```

#### **Arguments**

FALSE.

#### Value

A plot is returned

# **Examples**

```
categories_of_interest <- c("Tumour", "Immune1","Immune2","Immune3")
colour_vector <- c("red","darkblue","blue","darkgreen")
plot_cell_categories(SPIAT::defined_image, categories_of_interest, colour_vector,
feature_colname = "Cell.Type")</pre>
```

# **Description**

Plots distances between cells as a violin plot

#### Usage

```
plot_cell_distances_violin(cell_to_cell_dist)
```

#### **Arguments**

```
cell_to_cell_dist
```

Data.frame containing the distance output between cell types. The functions that generate the distances can be calculate\_minimum\_distances\_between\_celltypes and calculate\_pairwise\_distances\_between\_celltypes.

## Value

A plot is returned

```
distances <- calculate_pairwise_distances_between_celltypes(SPIAT::defined_image,
cell_types_of_interest = c("Immune1", "Immune2"), feature_colname="Cell.Type")
plot_cell_distances_violin(distances)</pre>
```

plot\_cell\_marker\_levels

# **Description**

Produces a scatter plot of the level of a marker in each cell. The level of the marker in all cells is shown, at x-y positions, no matter if cells are phenotyped as being positive or negative for the particular marker.

## Usage

```
plot_cell_marker_levels(spe_object, marker, feature_colname = "Phenotype")
```

# Arguments

```
spe_object SpatialExperiment object in the form of the output of format_image_to_spe.

marker String. Marker to plot.

feature_colname

String. Column containing marker information
```

## Value

A plot is returned

## **Examples**

```
plot_cell_marker_levels(SPIAT::simulated_image, "Immune_marker1")
```

```
plot_cell_percentages
```

# **Description**

Plots cells proportions as barplots.

```
plot_cell_percentages(
  cell_proportions,
  cells_to_exclude = NULL,
  cellprop_colname = "Proportion_name"
)
```

```
Cell_proportions

Data Frame. Output from calculate_cell_proportions.

cells_to_exclude

String Vector. Markers to exclude.

cellprop_colname

String. Column to use for y axis names. Default is "Proportion_name".
```

#### Value

A plot is returned

# **Examples**

```
plot_cell_percentages(p_cells)

plot_composition_heatmap
```

plot\_composition\_heatmap

p\_cells <- calculate\_cell\_proportions(SPIAT::simulated\_image)</pre>

# **Description**

Produces a heatmap showing the marker percentages within each cluster and the cluster sizes.

#### Usage

```
plot_composition_heatmap(
  composition,
  pheno_to_exclude = NULL,
  log_values = FALSE,
  feature_colname
)
```

## **Arguments**

#### Value

A plot is returned

plot\_distance\_heatmap

## **Examples**

```
neighborhoods <- identify_neighborhoods(image_no_markers, method = "hierarchical",
min_neighborhood_size = 100, cell_types_of_interest = c("Immune", "Immune1", "Immune2"),
radius = 50, feature_colname = "Cell.Type")
neighborhoods_vis <- composition_of_neighborhoods(neighborhoods, feature_colname="Cell.Type")
plot_composition_heatmap(neighborhoods_vis, feature_colname="Cell.Type")</pre>
```

plot\_distance\_heatmap

## **Description**

Takes the output of cell\_distances and plot the distances as a heatmap.

## Usage

```
plot_distance_heatmap(phenotype_distances_result, metric = "mean")
```

# Arguments

phenotype\_distances\_result

Dataframe output from 'calculate\_summary\_distances\_between\_celltypes' or 'calculate\_minimum\_distances\_between\_celltypes'.

metric

Metric to be plotted. One of "mean", "std.dev", "median", "min" or "max".

#### Value

A plot is returned

```
dists <- calculate_pairwise_distances_between_celltypes(SPIAT::defined_image,
  cell_types_of_interest = c("Tumour","Immune1"), feature_colname = "Cell.Type")
  summary_distances <- calculate_summary_distances_between_celltypes(dists)
  plot_distance_heatmap(summary_distances)</pre>
```

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```
plot_marker_level_heatmap

plot_marker_level_heatmap
```

## **Description**

Blurs the image by splitting the images into small squares. The marker levels are then averaged within each square. All cells are considered, regardless of phenotype status.

#### Usage

```
plot_marker_level_heatmap(spe_object, num_splits, marker)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

num\_splits Integer specifying the blurring level (number of splits) for the image. Higher numbers result in higher resolution.

marker String. Marker to plot.

## Value

A plot is returned

## **Examples**

```
plot_marker_level_heatmap(SPIAT::simulated_image, num_splits = 100, "Tumour_marker")
```

predict\_phenotypes

## **Description**

Predicts cell phenotypes based on marker intensity levels. If no prior cell phenotypes are available, it adds the phenotypes to the SpaitalExperiment object used as input. If reference cell phenotypes are available, it produces a density plot showing predicted cutoff of a positive reading for marker intensity and it returns a dataframe containing the predicted intensity status for a particular marker.

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#### Usage

```
predict_phenotypes(
    spe_object,
    thresholds = NULL,
    tumour_marker,
    baseline_markers,
    nuclear_marker = NULL,
    reference_phenotypes = FALSE,
    markers_to_phenotype = NULL,
    plot_distribution = TRUE
)
```

#### **Arguments**

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

thresholds (Optional) Numeric Vector specifying the cutoff of a positive reading. The order

must match the marker order, and it should be NA for DAPI.

tumour\_marker String containing the tumour\_marker used for the image. If tumor cells are

known, annotate tumor cells as 1 and non-tumor cells as 0, and include the

rowname.

baseline\_markers

String Vector. Markers not found on tumour cells to refine the threshold used for tumour cell phenotying.

nuclear\_marker String. Nuclear marker used.

reference\_phenotypes

Boolean. TRUE or FALSE value whether there are reference phenotypes for the sample obtained by the user through other means (e.g. HALO or InForm). If there are reference phenotypes available, a matrix of predicted phenotypes, intensities, and reference phenotypes will be returned, which can be used as input to "marker\_prediction\_plot". If no reference phenotype available, the result of the function will be added to the spe object used in the input. Note that if a reference phenotype is to be used, the phenotypes must be an explicit combination of positive markers (e.g. AMACR,PDL1), as opposed to descriptive (PDL1+tumour cells).

markers\_to\_phenotype

String Vector. Markers to be included in the phenotyping. If NULL, then all markers will be used. DAPI needs to be excluded.

plot\_distribution

Boolean. If TRUE, plots of the marker intensities distributions and cutoffs are plotted.

#### Value

An updated spe object with cell phenotypes or a data.frame of predicted phenotypes

 $S_{-}BC$ 

#### **Examples**

```
# keep the original phenotypes
predicted_result <- predict_phenotypes(spe_object = simulated_image, thresholds = NULL,
tumour_marker = "Tumour_marker",baseline_markers = c("Immune_marker1", "Immune_marker2",
"Immune_marker3", "Immune_marker4"), reference_phenotypes = TRUE)
# update the predicted phenotypes
predicted_spe_image <- predict_phenotypes(spe_object = simulated_image, thresholds = NULL,
tumour_marker = "Tumour_marker",baseline_markers = c("Immune_marker1", "Immune_marker2",
"Immune_marker3", "Immune_marker4"), reference_phenotypes = FALSE)</pre>
```

R\_BC

The ratio of border cell count to cluster cell count

#### **Description**

Calculates the ratio of the bordering cell count and the total to-be-clustered cell count in an image. The bordering cells are detected by the default identify\_bordering\_cells function. If the ratio is high, it means that most cells to be clustered are identified as bordering cells. This means there is no clear clusters.

## Usage

```
R_BC(spe_object, cell_type_of_interest, feature_colname)
```

## **Arguments**

#### Value

A number is returned.

```
R_BC(SPIAT::defined_image, cell_type_of_interest = "Tumour", "Cell.Type")
```

select\_celltypes 51

select\_celltypes select\_celltypes

# Description

Select cell types to keep or exclude in the analysis. The output of this function also includes the original image size and cell count.

## Usage

```
select_celltypes(
  spe_object,
  celltypes,
  feature_colname = "Phenotype",
  keep = TRUE
)
```

## Arguments

spe\_object SpatialExperiment object in the form of the output of format\_image\_to\_spe.

celltypes String Vector of celltypes of keep or exclude.

feature\_colname

String. The column that has the interested cell types. If the cells ids are used to

select cells, use "Cell.ID" for this arg.

keep Boolean. TRUE if vector of 'celltypes' are the cells that are going to be kept,

FALSE if they are to be removed.

## Value

A SpatialExperiment object is returned. The original image size and cell count can be accessed by 'attr(slim\_spe, "original\_cell\_number")' and 'attr(slim\_spe, "range\_of\_coords")', where 'slim\_spe' is the output of this function.

```
data_subset <- select_celltypes(SPIAT::simulated_image,
  celltypes = c("Tumour_marker","Immune_marker1","Immune_marker2",
  "Immune_marker3","Immune_marker4"),
  feature_colname = "Phenotype", keep=TRUE)
  attr(data_subset, "original_cell_number") #cell number in the original image
  attr(data_subset, "range_of_coords")
  dim(data_subset)[2] # this is the new image cell number</pre>
```

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simulated\_image

SPE object of a formatted image (simulated by 'spaSim' package)

# Description

A dataset that contains a formatted spe object with cell ids and phenotypes in 'colData()' and marker intensities in 'assays()'.

# Usage

simulated\_image

#### **Format**

An SpatialExperiment object. Assay contains 5 rows (markers) and 4951 columns (cells); colData contains 4951 rows (cells) and 3 columns.

## See Also

defined\_image image\_no\_markers

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