# Package 'CARDspa'

July 16, 2025

**Title** Spatially Informed Cell Type Deconvolution for Spatial Transcriptomics

**Version** 1.0.0 **Date** 2025-2-6

Description CARD is a reference-based deconvolution method that estimates cell type composition in spatial transcriptomics based on cell type specific expression information obtained from a reference scRNA-seq data. A key feature of CARD is its ability to accommodate spatial correlation in the cell type composition across tissue locations, enabling accurate and spatially informed cell type deconvolution as well as refined spatial map construction. CARD relies on an efficient optimization algorithm for constrained maximum likelihood estimation and is scalable to spatial transcriptomics with tens of thousands of spatial locations and tens of thousands of genes.

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assign_sc_cords	ion to assign the spatial location information for each single
-----------------	--

## Description

The function to assign the spatial location information for each single cell

#### Usage

```
assign_sc_cords(mappint_spot_cell_cor, cords_new, numcell, sc_eset, ct_varname)
```

## Arguments

mappint_spot_cell_cor		
		a mapped correlation matrix indicating the relashionship between each measured spatial location and the single cell in the scRNAseq reference
	cords_new	output from the function get_high_res_cords
	numcell	a numeric value indicating the number of single cells in each measured location, we suggest 20 for ST technology, 7 for 10x Viisum and 2 for Slide-seq
	sc_eset	a single cell experiment object stored in CARD object

character, the name of the column in metaData that specifies the cell type annotation information, stroed in CARD object

#### Value

ct\_varname

Return the assigned spatial location information for the mapped single cell

CADD along	Each CARD object has a number of alots which stone information. Von
CARD-class	Each CARD object has a number of slots which store information. Key
	slots to access are listed below.

## Description

Each CARD object has a number of slots which store information. Key slots to access are listed below.

#### Value

Return an object of CARD class

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#### **Slots**

sc\_eset The filtered scRNA-seq data along with meta data stored in the format of SingleCellExperiment.

spatial\_countMat The filtered spatial count data.

spatial\_location The weights for combining p-values from multiple kernels.

Proportion\_CARD The estimated cell type proportion by CARD with each row is a spatial location and each column is a cell type.

project The name of the project, default is deconvolution.

info\_parameters The paramters that are used in model fitting.

algorithm\_matrix The intermediate matrices that are used in the model fitting step.

refined\_prop The refined cell type proportion matrix estimated by CARD for the newly grided spatial locations. The number of initial grids are defined by the user.

refined\_expression The refined predicted expression matrix (normalized) estimated by CARD for the newly grided spatial locations. The number of initial grids are defined by the user.

CARDfree

SpatialDeconv function based on Conditional Autoregressive model

#### **Description**

SpatialDeconv function based on Conditional Autoregressive model

#### Usage

```
CARDfree(
   XinputIn,
   UIn,
   WIn,
   phiIn,
   max_iterIn,
   epsilonIn,
   initV,
   initb,
   initSigma_e2,
   initLambda
)
```

#### **Arguments**

XinputIn The input of normalized spatial data

UIn The input of cell type specific basis matrix B

WIn The constructed W weight matrix from Gaussian kernel

phiIn The phi value

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max\_iterIn Maximum iterations
epsilonIn epsilon for convergence

initV Initial matrix of cell type compositions V
initb Initial vector of cell type specific intercept

initSigma\_e2 Initial value of residual variance

initLambda Initial vector of cell type sepcific scalar.

#### Value

A list

CARDfree-class Each CARDfree object has a number of slots which store information.

Key slots to access are listed below.

#### **Description**

Each CARDfree object has a number of slots which store information. Key slots to access are listed below.

#### Value

Return an object of CARDfree class

#### **Slots**

spatial\_countMat The filtered spatial count data.

spatial\_location The weights for combining p-values from multiple kernels.

Proportion\_CARD The estimated cell type proportion by CARD with each row is a spatial location and each column is a cell type.

estimated\_refMatrix The estimated reference matrix by CARDfree with each row represents a gene and each column represents a cell type cluster.

project The name of the project, default is deconvolution.

markerList The nlist of cell type specific markers, with each element represents the vector of cell type specific markers

info\_parameters The paramters that are used in model fitting.

algorithm\_matrix The intermediate matrices that are used in the model fitting step.

refined\_prop The refined cell type proportion matrix estimated by CARD for the newly grided spatial locations. The number of initial grids are defined by the user.

refined\_expression The refined predicted expression matrix (normalized) estimated by CARD for the newly grided spatial locations. The number of initial grids are defined by the user.

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CARDref

SpatialDeconv function based on Conditional Autoregressive model

#### **Description**

SpatialDeconv function based on Conditional Autoregressive model

#### Usage

```
CARDref(
   XinputIn,
   UIn,
   WIn,
   phiIn,
   max_iterIn,
   epsilonIn,
   initV,
   initb,
   initSigma_e2,
   initLambda
)
```

#### **Arguments**

XinputIn The input of normalized spatial data

UIn The input of cell type specific basis matrix B

WIn The constructed W weight matrix from Gaussian kernel

phiIn The phi value

max\_iterIn Maximum iterations
epsilonIn epsilon for convergence

initV Initial matrix of cell type compositions V initb Initial vector of cell type specific intercept

initSigma\_e2 Initial value of residual variance

initLambda Initial vector of cell type sepcific scalar.

#### Value

A list

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 ${\it CARD\_deconvolution} \qquad {\it Spatially Informed Cell Type Deconvolution for Spatial Transcriptomics by CARD}$ 

## Description

Spatially Informed Cell Type Deconvolution for Spatial Transcriptomics by CARD

## Usage

```
CARD_deconvolution(
   sc_count,
   sc_meta,
   spatial_count,
   spatial_location,
   ct_varname,
   ct_select,
   sample_varname,
   mincountgene = 100,
   mincountspot = 5,
   sce = NULL,
   spe = NULL
)
```

## Arguments

sc_count	Raw scRNA-seq count data, each column is a cell and each row is a gene.
sc_meta	data frame, with each row representing the cell type and/or sample information of a specific cell. The row names of this data frame should match exactly with the column names of the $sc\_count$ data
spatial_count	Raw spatial resolved transcriptomics data, each column is a spatial location, and each row is a gene.
spatial_location	on
	data frame, with two columns representing the x and y coordinates of the spatial location. The rownames of this data frame should match eaxctly with the columns of the spatial_count.
ct_varname	character, the name of the column in metaData that specifies the cell type annotation information
ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. If NULL, then use all cell types provided by single cell dataset;
sample_varname	character,the name of the column in metaData that specifies the sample information. If NULL, we just use the whole as one sample.
mincountgene	Minimum counts for each gene
mincountspot	Minimum counts for each spatial location

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sce	a SingleCellExperiment object containing scRNA-seq count data in the counts assay, and cell types and sample information in the colData.
spe	a SpatialExperiment object containing spatial data in the counts assay, and spatial coordinates in the spatialCoords.

#### Value

Returns a SpatialExperiment object with estimated cell type proportion stored in object\$Proportion\_CARD.

#### **Examples**

```
library(RcppML)
library(NMF)
library(RcppArmadillo)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
    ct_select = unique(sc_meta$cellType),
    sample_varname = "sampleInfo",
    mincountgene = 100,
    mincountspot = 5
)
```

CARD\_imputation

Construct an enhanced spatial expression map on the unmeasured tissue locations

#### **Description**

Construct an enhanced spatial expression map on the unmeasured tissue locations

## Usage

```
CARD_imputation(CARD_object, num_grids, ineibor = 10, exclude = NULL)
```

## Arguments

CARD\_object SpatialExperiment Object created by CARD\_deconvolution with estimated cell

type compositions on the original spatial resolved transcriptomics data.

num\_grids Initial number of newly grided spatial locations. The final number of newly

grided spatial locations will be lower than this value since the newly grided

locations outside the shape of the tissue will be filtered

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ineibor Numeric, number of neighbors used in the imputation on newly grided spatial

locations, default is 10.

exclude Vector, the rownames of spatial location data on the original resolution that you

want to exclude. This is to avoid the weird detection of the shape.

#### Value

Return a SpatialExperiment object with the refined cell type compositions estimated for newly grided spots and the refined predicted gene expression (normalized).

## Examples

```
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
    ct_select = unique(sc_meta$cellType),
    sample_varname = "sampleInfo",
   mincountgene = 100,
   mincountspot = 5
CARD_obj <- CARD_imputation(</pre>
    CARD_obj,
    num_grids = 200,
    ineibor = 10,
    exclude = NULL
)
```

CARD\_refFree

Extension of CARD into a reference-free version of deconvolution: CARDfree.

## Description

Extension of CARD into a reference-free version of deconvolution: CARDfree.

## Usage

```
CARD_refFree(
  markerlist,
  spatial_count,
  spatial_location,
```

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```
mincountgene = 100,
mincountspot = 5,
spe = NULL
)
```

#### **Arguments**

markerlist a list of marker genes, with each element of the list being the vector of cell type

specific marker genes

spatial\_count Raw spatial resolved transcriptomics data, each column is a spatial location, and

each row is a gene.

spatial\_location

data frame, with two columns representing the x and y coordinates of the spatial location. The rownames of this data frame should match eaxctly with the

columns of the spatial\_count.

mincountgene Minimum counts for each gene

mincountspot Minimum counts for each spatial location

spe a SpatialExperiment object containing spatial data in the counts assay, and

spatial coordinates in the spatialCoords.

#### Value

Returns a SpatialExperiment object with estimated cell type proportion stored in object\$Proportion\_CARD. Because this is a reference-free version, the columns of estimated proportion is not cell type but cell type cluster

```
library(RcppML)
library(NMF)
library(RcppArmadillo)
data(markerList)
data(spatial_count)
data(spatial_location)
CARDfree_obj <- CARD_refFree(
markerlist = markerList[8:16],
spatial_count = spatial_count[1:2500, ],
spatial_location = spatial_location,
mincountgene = 100,
mincountspot = 5
)</pre>
```

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CARD_scmapping	Extension of CARD into performing single cell Mapping from non- single cell spatial transcriptomics dataset.

## **Description**

Extension of CARD into performing single cell Mapping from non-single cell spatial transcriptomics dataset.

#### Usage

```
CARD_scmapping(CARD_object, shapeSpot = "Square", numcell, ncore = 10)
```

## Arguments

CARD_object	CARD object create by the CARD_deconvolution function.
shapeSpot	a character indicating whether the sampled spatial coordinates for single cells locating in a Square-like region or a Circle-like region. The center of this region is the measured spatial location in the non-single cell resolution spatial transcriptomics data. The default is 'Square', the other shape is 'Circle'
numcell	a numeric value indicating the number of single cells in each measured location, we suggest 20 for ST technology, 7 for 10x Viisum and 2 for Slide-seq
ncore	a numeric value indicating the number of cores used to accelerating the procedure

#### Value

Returns a SingleCellExperiment SCE object with the mapped expression at single cell resolution and the spatial location information of each single cell

```
library(SingleCellExperiment)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
   sc_count = sc_count,
   sc_meta = sc_meta,
   spatial_count = spatial_count,
    spatial_location = spatial_location,
   ct_varname = "cellType",
   ct_select = unique(sc_meta$cellType),
   sample_varname = "sampleInfo",
   mincountgene = 100,
   mincountspot = 5
)
```

CARD\_visualize\_Cor

```
scMapping <- CARD_scmapping(
CARD_obj,
shapeSpot = "Square",
numcell = 20,
ncore = 2)
print(scMapping)</pre>
```

CARD\_visualize\_Cor

Visualize the cell type proportion correlation

## Description

Visualize the cell type proportion correlation

#### Usage

```
CARD_visualize_Cor(proportion, colors = colors)
```

#### **Arguments**

proportion Data frame, cell type proportion estimated by CARD in either original resolution

or enhanced resolution.

colors Vector of color names that you want to use, if NULL, we will use the default

color scale c("#91a28c","white","#8f2c37")

#### Value

Returns a ggcorrplot figure.

```
library(ggplot2)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
   sc_count = sc_count,
   sc_meta = sc_meta,
   spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
   ct_select = unique(sc_meta$cellType),
    sample_varname = "sampleInfo",
   mincountgene = 100,
   mincountspot = 5
CARD_visualize_Cor(CARD_obj$Proportion_CARD, colors = NULL)
```

CARD\_visualize\_gene

Visualize the spatial distribution of cell type proportion

#### **Description**

Visualize the spatial distribution of cell type proportion

#### Usage

```
CARD_visualize_gene(
   spatial_expression,
   spatial_location,
   gene_visualize,
   colors = colors,
   NumCols
)
```

#### **Arguments**

spatial\_expression

Data frame, spatial gene expression in either original resolution or enhanced

resolution.

spatial\_location

Data frame, spatial location information.

gene\_visualize Vector of selected gene names that are interested to visualize

colors Vector of color names that you want to use, if NULL, we will use the default

color scale in virdis palette

NumCols Numeric, number of columns in the figure panel, it depends on the number of

cell types you want to visualize.

#### Value

Returns a ggplot2 figure.

```
library(ggplot2)
library(SummarizedExperiment)
library(SpatialExperiment)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,</pre>
```

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```
spatial_location = spatial_location,
  ct_varname = "cellType",
  ct_select = unique(sc_meta$cellType),
  sample_varname = "sampleInfo",
  mincountgene = 100,
  mincountspot = 5
)

CARD_visualize_gene(
  spatial_expression = assays(CARD_obj)$spatial_countMat,
  spatial_location = spatialCoords(CARD_obj),
  gene_visualize = c("A4GNT", "AAMDC", "CD248"),
  colors = NULL,
  NumCols = 3
)
```

CARD\_visualize\_pie

Visualize the spatial distribution of cell type proportion in a geom scatterpie plot

#### **Description**

Visualize the spatial distribution of cell type proportion in a geom scatterpie plot

#### Usage

```
CARD_visualize_pie(proportion, spatial_location, colors = NULL, radius = NULL)
```

#### **Arguments**

proportion Data frame, cell type proportion estimated by CARD in either original resolution

or enhanced resolution.

spatial\_location

Data frame, spatial location information.

colors Vector of color names that you want to use, if NULL, we will use the color

palette "Spectral" from RColorBrewer package.

radius Numeric value about the radius of each pie chart, if NULL, we will calculate it

inside the function.

#### Value

Returns a ggplot2 figure.

#### **Examples**

```
library(ggplot2)
library(SpatialExperiment)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
    ct_select = unique(sc_meta$cellType),
    sample_varname = "sampleInfo",
    mincountgene = 100,
    mincountspot = 5
)
colors <- c(
    "#FFD92F", "#4DAF4A", "#FCCDE5", "#D9D9D9", "#377EB8", "#7FC97F",
    "#BEAED4", "#FDC086", "#FFFF99", "#386CB0", "#F0027F", "#BF5B17", "#666666", "#1B9E77", "#D95F02", "#7570B3", "#E7298A", "#66A61E",
    "#E6AB02", "#A6761D"
)
CARD_visualize_pie(
    proportion = CARD_obj$Proportion_CARD,
    spatial_location = spatialCoords(CARD_obj),
    colors = colors,
    radius = 0.52
)
```

CARD\_visualize\_prop

Visualize the spatial distribution of cell type proportion

#### **Description**

Visualize the spatial distribution of cell type proportion

## Usage

```
CARD_visualize_prop(
  proportion,
  spatial_location,
  ct_visualize = ct_visualize,
  colors = c("lightblue", "lightyellow", "red"),
  NumCols,
  pointSize = 3
)
```

#### **Arguments**

proportion Data frame, cell type proportion estimated by CARD in either original resolution or enhanced resolution.

spatial\_location
Data frame, spatial location information.

ct\_visualize Vector of selected cell type names that are interested to visualize

colors Vector of color names that you want to use, if NULL, we will use the default color scale c("lightblue","lightyellow","red")

NumCols Numeric, number of columns in the figure panel, it depends on the number of cell types you want to visualize.

pointSize Size of each point used for plotting

#### Value

Returns a ggplot2 figure.

```
library(ggplot2)
library(SpatialExperiment)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(</pre>
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
    ct_select = unique(sc_meta$cellType),
    sample_varname = "sampleInfo",
   mincountgene = 100,
   mincountspot = 5
ct_visualize <- c(</pre>
    "Acinar_cells", "Cancer_clone_A", "Cancer_clone_B",
    "Ductal_terminal_ductal_like", "Ductal_CRISP3_high-centroacinar_like",
    "Ductal_MHC_Class_II", "Ductal_APOL1_high-hypoxic", "Fibroblasts"
CARD_visualize_prop(
    proportion = CARD_obj$Proportion_CARD,
    spatial_location = spatialCoords(CARD_obj),
    ct_visualize = ct_visualize,
    colors = c("lightblue", "lightyellow", "red"),
   NumCols = 4,
    pointSize = 3.0
)
```

```
CARD_visualize_prop_2CT
```

Visualize the spatial distribution of two cell type proportions on the same plot

#### **Description**

Visualize the spatial distribution of two cell type proportions on the same plot

#### Usage

```
CARD_visualize_prop_2CT(
  proportion,
  spatial_location,
  ct2_visualize = ct2_visualize,
  colors = NULL
)
```

#### **Arguments**

proportion Data frame, cell type proportion estimated by CARD in either original resolution

or enhanced resolution.

spatial\_location

Data frame, spatial location information.

ct2\_visualize Vector of selected two cell type names that are interested to visualize, here we

only focus on two cell types

colors list of color names that you want to use for each cell type, if NULL, we will use

the default color scale list list(c("lightblue", "lightyellow", "red"), c("lightblue", "lightyellow", "black")

#### Value

Returns a ggplot2 figure.

```
library(ggplot2)
library(SpatialExperiment)
data(spatial_count)
data(spatial_location)
data(sc_count)
data(sc_meta)
CARD_obj <- CARD_deconvolution(
    sc_count = sc_count,
    sc_meta = sc_meta,
    spatial_count = spatial_count,
    spatial_location = spatial_location,
    ct_varname = "cellType",
    ct_select = unique(sc_meta$cellType),</pre>
```

#### **Description**

Create the CARD object

#### Usage

```
createCARDfreeObject(
  markerlist,
  spatial_count,
  spatial_location,
  mincountgene = 100,
  mincountspot = 5,
  spe = NULL
)
```

#### **Arguments**

markerlist a list of marker genes, with each element of the list being the vector of cell type

specific marker genes

spatial\_count Raw spatial resolved transcriptomics data, each column is a spatial location, and

each row is a gene.

spatial\_location

data frame, with two columns representing the x and y coordinates of the spatial location. The rownames of this data frame should match eaxctly with the

columns of the spatial\_count.

mincountgene Minimum counts for each gene

mincountspot Minimum counts for each spatial location

spe a SpatialExperiment object containing spatial data in the counts assay, and

spatial coordinates in the spatial Coords.

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## Value

Returns CARDfree object with filtered spatial count and marker gene list.

createCARDObject

Create the CARD object

## Description

Create the CARD object

## Usage

```
createCARDObject(
   sc_count,
   sc_meta,
   spatial_count,
   spatial_location,
   ct_varname,
   ct_select,
   sample_varname,
   mincountgene = 100,
   mincountspot = 5,
   sce = NULL,
   spe = NULL
)
```

#### **Arguments**

sc_count	Raw scRNA-seq count data, each column is a cell and each row is a gene.
sc_meta	data frame, with each row representing the cell type and/or sample information of a specific cell. The row names of this data frame should match exactly with the column names of the $sc\_count$ data
spatial_count	Raw spatial resolved transcriptomics data, each column is a spatial location, and each row is a gene.
spatial_location	on .
	data frame, with two columns representing the x and y coordinates of the spatial location. The rownames of this data frame should match eaxctly with the columns of the spatial_count.
ct_varname	character, the name of the column in metadata that specifies the cell type annotation information
ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. If NULL, then use all cell types provided by single cell dataset;
sample_varname	character, the name of the column in metadata that specifies the sample information. If NULL, we just use the whole as one sample.
mincountgene	Minimum counts for each gene

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mincountspot	Minimum counts for each spatial location
sce	a SingleCellExperiment object containing scRNA-seq count data in the counts assay, and cell types and sample information in the colData.
spe	a SpatialExperiment object containing spatial data in the counts assay, and spatial coordinates in the spatialCoords.

#### Value

Returns CARD object with filtered spatial count and single cell RNA-seq dataset.

create_ref Construct the mean gene expression basis matrix (B), this is the faster version	?r
--	----

## Description

Construct the mean gene expression basis matrix (B), this is the faster version

## Usage

```
create_ref(sc_eset, ct_select = NULL, ct_varname, sample_varname = NULL)
```

## Arguments

sc_eset	S4 class for storing data from single-cell experiments. This format is usually created by the package SingleCellExperiment with stored counts, along with the usual metadata for genes and cells.
ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. If NULL, then use all cell types provided by single cell dataset;
ct_varname	character, the name of the column in metaData that specifies the cell type annotation information
sample_varname	character, the name of the column in metaData that specifies the sample information. If NULL, we just use the whole as one sample.

## Value

Return a list of basis (B) matrix

get\_high\_res\_cords 21

get_high_res_cords The cell	function to sample the spatial location information for each single
-----------------------------	---

## Description

The function to sample the spatial location information for each single cell

#### Usage

```
get_high_res_cords(cords, numcell, shape = "Square")
```

## Arguments

cords	The spatial location information in the measure spatial locations, with the first and second columns represent the 2-D x-y coordinate system
numcell	a numeric value indicating the number of single cells in each measured location, we suggest 20 for ST technology, 7 for 10x Viisum and 2 for Slide-seq
shape	a character indicating whether the sampled spatial coordinates for single cells locating in a Square-like region or a Circle-like region. The center of this region is the measured spatial location in the non-single cell resolution spatial transcriptomics data. The default is 'Square', the other shape is 'Circle'

#### Value

Returns a dataframe with the sampled spatial location information for each single cell

get_weight_for_cell
---------------------

## Description

The function to estimate the cell type composition signature for each single cell in the scRNaseq reference data

## Usage

```
get_weight_for_cell(sc_eset, ct_varname, ct_select, sample_varname, B)
```

22 mvn\_cv

## **Arguments**

tation information, stored in the CARD object	sc_eset	the sc_eset stored in the CARD object
NULL. stored in the CARD object sample_varname character,the name of the column in metaData that specifies the sample information. stored in the CARD object	ct_varname	character, the name of the column in metaData that specifies the cell type annotation information, stored in the CARD object
mation. stored in the CARD object	ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. stored in the CARD object
B reference basis matrix stored in the CARD object.	sample_varname	· · · · · · · · · · · · · · · · · · ·
	В	reference basis matrix stored in the CARD object.

#### Value

Returns a matrix of the cell type composition signature for each single cell in the scRNaseq reference

|--|

## Description

The marker gene list is a list format with each element of the list being the cell type specific gene markers.

#### Usage

```
data(markerList)
```

#### **Format**

An object of class list of length 20.

mvn_cv	Imputation and Construction of High-Resolution Spatial Maps for Cell Type Composition and Gene Expression by the spatial correlation structure between original spatial locations and new grided spatial locations
mvn_cv	Type Composition and Gene Expression by the spatial correlation structure between original spatial locations and new grided spatial

## Description

Imputation and Construction of High-Resolution Spatial Maps for Cell Type Composition and Gene Expression by the spatial correlation structure between original spatial locations and new grided spatial locations

#### Usage

```
mvn_cv(
  vtrain,
  location_orig,
  train_ind,
  test_ind,
  B,
  xinput_norm,
  optimal_b,
  optimal_phi,
  lambda,
  ineibor
)
```

## Arguments

vtrain	Matrix, estimated V matrix from CARD
location_orig	Data frame, spatial location data frame of the original spatial resolved transcriptomics dataset, stored in the spatialCoords(CARD_object)
train_ind	Vector, index of the original spatial locations
test_ind	Vector, index of the newly grided spatial locations
В	Matrix, used in the deconvolution as the reference basis matrix
xinput_norm	Matrix, used in the deconvolution as the normalized spatial count data
optimal_b	Vector, vector of the intercept for each cel type estimated based on the original spatial resolution
optimal_phi	Numeric, the optimal phi value stored in CARD_object
lambda	Vector, vector of cell type specific scalar in the CAR model
ineibor	Numeric, number of neighbors used in the imputation on newly grided spatial locations, default is 10.

#### Value

Return a list with the imputed Cell type composition Vtest matrix on the newly grided spatial locations and predicted normalized gene expression

```
norm_coords_train_test
```

Normalize the new spatial locations without changing the shape and relative positions

## Description

Normalize the new spatial locations without changing the shape and relative positions

24 sample\_grid\_within

#### Usage

```
norm_coords_train_test(location_orig, train_ind, test_ind)
```

#### **Arguments**

location\_orig Data frame, spatial location data frame of the original spatial resolved transcrip-

tomics dataset, stored in the spatialCoords(CARD\_object)

train\_ind Vector, Index of the original spatial locations

test\_ind Vector, Index of the newly grided spatial locations

#### Value

Return the normalized spatial location data frame

sample\_grid\_within Make new spatial locations on unmeasured tissue through grids.

#### **Description**

Make new spatial locations on unmeasured tissue through grids.

## Usage

```
sample_grid_within(location, num_sample, concavity = 2)
```

## **Arguments**

location Data frame, spatial location data frame of the original spatial resolved transcrip-

tomics dataset, stored in the spatialCoords(CARD\_object)

num\_sample Numeric, approximate number of cells in grid within the shape of the spatial

location data frame

concavity Numeric, a relative measure of concavity. The default is 2.0, which can prode-

cure detailed enough shapes. Infinity results in a convex hull while 1 results in

a more detailed shape.

#### Value

Return a list of data frame with newly grided points

sc\_count 25

sc\_count

scRNA-seq count data

#### **Description**

The scRNA-seq count data must be in the format of matrix or sparseMatrix, while each row represents a gene and each column represents a cell.

#### Usage

```
data(sc_count)
```

#### **Format**

An object of class dgCMatrix with 7000 rows and 1926 columns.

sc\_meta

scRNAseq meta data

#### **Description**

The scRNAseq meta data must be in the format of data frame while each row represents a cell. The rownames of the scRNAseq meta data should match exactly with the column names of the scRNAseq count data. The sc\_meta data must contain the column indicating the cell type assignment for each cell (e.g., "cellType" column in the example sc\_meta data). Sample/subject information should be provided, if there is only one sample, we can add a column by sc\_meta\$sampleInfo = "sample1".

#### Usage

```
data(sc_meta)
```

#### **Format**

An object of class data. frame with 1926 rows and 3 columns.

26 sc\_QC

sc\_QC

Quality control of scRNA-seq count data

## Description

Quality control of scRNA-seq count data

## Usage

```
sc_QC(
  counts_in,
  metadata,
  ct_varname,
  ct_select,
  sample_varname = NULL,
  min_cells = 0,
  min_genes = 0
)
```

## Arguments

counts_in	Raw scRNAseq count data, each column is a cell and each row is a gene.
metadata	data frame, metadata with "ct_varname" specify the cell type annotation information and "sample_varname" specify the sample information
ct_varname	character, the name of the column in metadata that specifies the cell type annotation information
ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. If NULL, then use all cell types provided by single cell dataset;
sample_varname	character, the name of the column in metadata that specifies the sample information. If NULL, we just use the whole as one sample.
min_cells	numeric, we filtered out the non-expressed cells.
min_genes	numeric we filtered out the non-expressed genes

## Value

Return the filtered scRNA-seq data and meta data stored in a S4 class (SingleCellExperiment)

select\_info 27

select_info Select Informative Genes used in the deconvolution	select_info	Select Informative Genes used in the deconvolution	
--	-------------	--	--

## Description

Select Informative Genes used in the deconvolution

#### Usage

```
select_info(basis, sc_eset, commongene, ct_select, ct_varname)
```

## Arguments

basis	Reference basis matrix.
sc_eset	scRNAseq data along with meta data stored in the S4 class format (SingleCell-Experiment).
commongene	common genes between scRNAseq count data and spatial resolved transcriptomics data.
ct_select	vector of cell type names that you are interested in to deconvolute, default as NULL. If NULL, then use all cell types provided by single cell dataset;
ct_varname	character, the name of the column in metaData that specifies the cell type annotation information

#### Value

a vector of informative genes selected

## Description

This method provides a concise summary of an object of class CARD, displaying key information including the project name, the number of spots, the number of cell types, and a sample of the Proportion\_CARD matrix.

## Usage

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

## Arguments

object An object of class CARD.

28 Sigma

#### Value

A concise summary of the CARD object is printed to the console.

show, CARDfree-method Show method for the CARDfree class

#### **Description**

This method provides a concise summary of an object of class CARDfree, displaying key information including the project name, the number of spots, the number of cell types, and a sample of the Proportion\_CARD matrix.

## Usage

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

#### Arguments

object An object of class CARDfree.

#### Value

A concise summary of the CARDfree object is printed to the console.

Sigma Calculate the variance covariance matrix used in the imputation of the new grided locations

#### **Description**

Calculate the variance covariance matrix used in the imputation of the new grided locations

#### Usage

```
Sigma(location_orig, train_ind, test_ind, optimal_phi, ineibor)
```

#### **Arguments**

location_orig	Data frame, spatial location data frame of the original spatial resolved transcriptomics dataset, stored in the spatialCoords(CARD_object)
train_ind	Vector, index of the original spatial locations
test_ind	Vector, index of the newly grided spatial locations
optimal_phi	Numeric, the optimal phi value stored in CARD_object
ineibor	Numeric, number of neighbors used in the imputation on newly grided spatial locations, default is 10.

spatial\_count 29

#### Value

Return a list with the imputed Cell type composition Vtest matrix on the newly grided spatial locations and predicted normalized gene expression

spatial\_count

Spatial transcriptomics count data

#### **Description**

The spatial transcriptomics count data must be in the format of matrix or sparseMatrix, while each row represents a gene and each column represents a spatial location. The column names of the spatial data can be in the "XcoordxYcoord" (i.e., 10x10) format, but you can also maintain your original spot names, for example, barcode names.

#### Usage

```
data(spatial_count)
```

#### **Format**

An object of class dgCMatrix with 11000 rows and 428 columns.

spatial\_location

Spatial location data

#### **Description**

The spatial location data must be in the format of data frame while each row represents a spatial location, the first column represents the x coordinate and the second column represents the y coordinate. The rownames of the spatial location data frame should match exactly with the column names of the spatial\_count.

#### Usage

```
data(spatial_location)
```

#### **Format**

An object of class data. frame with 428 rows and 2 columns.

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