

# Package ‘corral’

April 3, 2025

**Title** Correspondence Analysis for Single Cell Data

**Version** 1.17.0

**Date** 2023-02-09

**Description** Correspondence analysis (CA) is a matrix factorization method, and is similar to principal components analysis (PCA). Whereas PCA is designed for application to continuous, approximately normally distributed data, CA is appropriate for non-negative, count-based data that are in the same additive scale. The corral package implements CA for dimensionality reduction of a single matrix of single-cell data, as well as a multi-table adaptation of CA that leverages data-optimized scaling to align data generated from different sequencing platforms by projecting into a shared latent space. corral utilizes sparse matrices and a fast implementation of SVD, and can be called directly on Bioconductor objects (e.g., SingleCellExperiment) for easy pipeline integration. The package also includes additional options, including variations of CA to address overdispersion in count data (e.g., Freeman-Tukey chi-squared residual), as well as the option to apply CA-style processing to continuous data (e.g., proteomic TOF intensities) with the Hellinger distance adaptation of CA.

**Imports** ggplot2, ggthemes, grDevices, gridExtra, irlba, Matrix, methods, MultiAssayExperiment, pals, reshape2, SingleCellExperiment, SummarizedExperiment, transport

**Suggests** ade4, BiocStyle, CellBench, DuoClustering2018, knitr, rmarkdown, scater, testthat

**License** GPL-2

**RoxygenNote** 7.1.2

**VignetteBuilder** knitr

**biocViews** BatchEffect, DimensionReduction, GeneExpression, Preprocessing, PrincipalComponent, Sequencing, SingleCell, Software, Visualization

**Encoding** UTF-8

**git\_url** <https://git.bioconductor.org/packages/corral>

**git\_branch** devel

**git\_last\_commit** 284ef14

**git\_last\_commit\_date** 2024-10-29

**Repository** Bioconductor 3.21

**Date/Publication** 2025-04-03

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## Contents

add_embeddings2scelist . . . . .	2
all_are . . . . .	3
biplot_corral . . . . .	4
compsvd . . . . .	5
corralm_matlist . . . . .	6
corral_mat . . . . .	8
corral_preproc . . . . .	11
earthmover_dist . . . . .	12
get_pct_var_exp_svd . . . . .	13
get_weights . . . . .	14
list2mat . . . . .	14
na2zero . . . . .	15
obs2probs . . . . .	15
pairwise_rv . . . . .	16
plot_embedding . . . . .	17
plot_embedding_sce . . . . .	18
rv . . . . .	20
scal_var . . . . .	21
scal_var_mat . . . . .	22
sce2matlist . . . . .	22
trim_matdist . . . . .	23
var_stabilize . . . . .	24
<b>Index</b>	<b>25</b>

---

add\_embeddings2scelist

*Add embeddings to list of SCEs*

---

### Description

Add embeddings to list of SCEs

### Usage

```
add_embeddings2scelist(scelist, embeddings, slotname = "corralm")
```

**Arguments**

sclist	list of SingleCellExperiments; to which the corresponding embeddings should be added
embeddings	matrix; the embeddings outputted from a dimension reduction, e.g. <code>corralm</code> . Rows in this table correspond to columns in the SCEs in <code>sclist</code> (if all the SCEs were column-bound), and row indices should correspond to cells.
slotname	character; name of the slot for the reduced dim embedding; defaults to <code>corrall</code>

**Value**

list of SingleCellExperiments with respective embeddings stored in them

**Examples**

```
library(DuoClustering2018)
sce <- sce_full_Zhengmix4eq()
sclist <- list(sce,sce)
embeddings <- matrix(sample(seq(0,20,1),dim(sce)[2]*6,replace = TRUE),nrow = dim(sce)[2]*2)
sclist <- add_embeddings2sclist(sclist, embeddings)
```

---

all_are	<i>all_are</i>
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---

**Description**

Checks if all elements of a list or List are of a (single) particular type `typechar`

**Usage**

```
all_are(inplist, typechar)
```

**Arguments**

inplist	list or List to be checked
typechar	char of the type to check for

**Value**

boolean, for whether the elements of `inplist` are all `typechar`

**Examples**

```
x <- list(1,2)
all_are(x,'numeric')
all_are(x,'char')

y <- list(1,2,'c')
all_are(y,'numeric')
all_are(y,'char')
```

---

<code>biplot_corrall</code>	<i>Generate biplot for corral object</i>
-----------------------------	--

---

**Description**

Generate biplot for corral object

**Usage**

```
biplot_corrall(
  corral_obj,
  color_vec,
  text_vec,
  feat_name = "(genes)",
  nfeat = 20,
  xpc = 1,
  plot_title = "Biplot",
  text_size = 2,
  xjitter = 0.005,
  yjitter = 0.005,
  coords = c("svd", "PC", "SC")
)
```

**Arguments**

<code>corral_obj</code>	list outputted by the <code>corral</code> function
<code>color_vec</code>	vector; length should correspond to the number of rows in <code>v</code> of <code>corral_obj</code> , and each element of the vector classifies that cell (entry) in the embedding to that particular class, which will be colored the same. (e.g., cell type)
<code>text_vec</code>	vector; length should correspond to the number of rows in <code>u</code> of <code>corral_obj</code> , and each element of the vector is the label for the respective feature that would show on the biplot.
<code>feat_name</code>	char; the label will in the legend. Defaults to <code>(genes)</code> .
<code>nfeat</code>	int; the number of features to include. The function will first order them by distance from origin in the selected dimensions, then select the top <code>n</code> to be displayed.
<code>xpc</code>	int; which PC to put on the x-axis (defaults to 1)
<code>plot_title</code>	char; title of plot (defaults to <code>*Biplot*</code> )
<code>text_size</code>	numeric; size of the feature labels given in <code>text_vec</code> (defaults to 2; for <code>ggplot2</code> )
<code>xjitter</code>	numeric; the amount of jitter for the text labels in x direction (defaults to .005; for <code>ggplot2</code> )
<code>yjitter</code>	numeric; the amount of jitter for the text labels in y direction (defaults to .005; for <code>ggplot2</code> )
<code>coords</code>	char; indicator for sets of coordinates to use. <code>svd</code> plots the left and right singular vectors as outputted by SVD ( <code>u</code> and <code>v</code> ), which <code>PC</code> and <code>SC</code> use the principal and standard coordinates, respectively (defaults to <code>svd</code> )

**Value**

ggplot2 object of the biplot

**Examples**

```
library(DuoClustering2018)
library(SingleCellExperiment)
zm4eq.sce <- sce_full_Zhengmix4eq()
zm4eq.countmat <- counts(zm4eq.sce)
zm4eq.corral_obj <- corral(zm4eq.countmat)
gene_names <- rowData(zm4eq.sce)$symbol
ctvec <- zm4eq.sce$phenoid

biplot_corral(corral_obj = zm4eq.corral_obj, color_vec = ctvec, text_vec = gene_names)
```

---

compsvd

*compsvd: Compute Singular Value Decomposition (SVD)*


---

**Description**

Computes SVD.

**Usage**

```
compsvd(mat, method = c("irl", "svd"), ncomp = 30, ...)
```

**Arguments**

mat	matrix, pre-processed input; can be sparse or full (pre-processing can be performed using <a href="#">corral_preproc</a> from this package)
method	character, the algorithm to be used for svd. Default is irl. Currently supports 'irl' for irlba::irlba or 'svd' for stats::svd
ncomp	numeric, number of components; Default is 30
...	(additional arguments for methods)

**Value**

SVD result - a list with the following elements:

d a vector of the diagonal singular values of the input mat. Note that using svd will result in the full set of singular values, while irlba will only compute the first ncomp singular values.

u a matrix of with the left singular vectors of mat in the columns

v a matrix of with the right singular vectors of mat in the columns

eigsum sum of the eigenvalues, for calculating percent variance explained

**Examples**

```
mat <- matrix(sample(0:10, 2500, replace=TRUE), ncol=50)
compsvd(mat, method = 'irl', ncomp = 5)
```

---

corralm\_matlist      *Multi-table correspondence analysis (list of matrices)*

---

### Description

This multi-table adaptation of correspondence analysis applies the same scaling technique and enables data alignment by finding a set of embeddings for each dataset within shared latent space.

### Usage

```
corralm_matlist(
  matlist,
  method = c("irl", "svd"),
  ncomp = 30,
  rtype = c("indexed", "standardized", "hellinger", "freemantukey", "pearson"),
  vst_mth = c("none", "sqrt", "freemantukey", "anscombe"),
  rw_contrib = NULL,
  ...
)

corralm_sce(
  sce,
  splitby,
  method = c("irl", "svd"),
  ncomp = 30,
  whichmat = "counts",
  fullout = FALSE,
  rw_contrib = NULL,
  ...
)

corralm(inp, whichmat = "counts", fullout = FALSE, ...)

## S3 method for class 'corralm'
print(x, ...)
```

### Arguments

matlist	(for corralm_matlist) list of input matrices; input matrices should be counts (raw or log). Matrices should be aligned row-wise by common features (either by sample or by gene)
method	character, the algorithm to be used for svd. Default is irl. Currently supports 'irl' for irlba::irlba or 'svd' for stats::svd
ncomp	numeric, number of components; Default is 30
rtype	character indicating what type of residual should be computed; options are "indexed", "standardized" (or "pearson" is equivalent), "freemantukey",

	and "hellinger"; defaults to "standardized" for <code>corral</code> and "indexed" for <code>corralm</code> . "indexed", "standardized", and "freemantukey" compute the respective chi-squared residuals and are appropriate for count data. The "hellinger" option is appropriate for continuous data.
<code>vst_mth</code>	character indicating whether a variance-stabilizing transform should be applied prior to calculating chi-squared residuals; defaults to "none"
<code>rw_contrib</code>	numeric vector, same length as the matlist. Indicates the weight that each dataset should contribute to the row weights. When set to NULL the row weights are *not* combined and each matrix is scaled independently (i.e., using their observed row weights, respectively). When set to a vector of all the same values, this is equivalent to taking the mean. Another option is to the number of observations per matrix to create a weighted mean. Regardless of input scale, row weights for each table must sum to 1 and thus are scaled. When this option is specified (i.e., not 'NULL'), the 'rtype' argument will automatically be set to 'standardized', and whatever argument is given will be ignored.
<code>...</code>	(additional arguments for methods)
<code>sce</code>	(for <code>corralm_sce</code> ) SingleCellExperiment; containing the data to be integrated. Default is to use the counts, and to include all of the data in the integration. These can be changed by passing additional arguments. See <code>sce2matlist</code> function documentation for list of available parameters.
<code>splitby</code>	character; name of the attribute from <code>colData</code> that should be used to separate the SCE.
<code>whichmat</code>	char, when using SingleCellExperiment or other SummarizedExperiment, can be specified. default is 'counts'.
<code>fullout</code>	boolean; whether the function will return the full <code>corralm</code> output as a list, or a SingleCellExperiment; defaults to SingleCellExperiment (FALSE). To get back the <code>corralm_matlist</code> -style output, set this to TRUE.
<code>inp</code>	list of matrices (any type), a SingleCellExperiment, list of SingleCellExperiments, list of SummarizedExperiments, or MultiAssayExperiment. If using SingleCellExperiment or SummarizedExperiment, then include the <code>whichmat</code> argument to specify which slot to use (defaults to counts). Additionally, if it is one SingleCellExperiment, then it is also necessary to include the <code>splitby</code> argument to specify the batches. For a MultiAssayExperiment, it will take the intersect of the features across all the assays, and use those to match the matrices; to use a different subset, select desired subsets then call <code>corral</code>
<code>x</code>	(print method) <code>corralm</code> object; the list output from <code>corralm_matlist</code>

## Details

`corralm` is a wrapper for `corralm_matlist` and `corralm_sce`, and can be called on any of the acceptable input types (see `inp` below).

## Value

When run on a list of matrices, a list with the correspondence analysis matrix decomposition result, with indices corresponding to the concatenated matrices (in order of the list):

**d** a vector of the diagonal singular values of the input mat (from SVD output)  
**u** a matrix of with the left singular vectors of mat in the columns (from SVD output)  
**v** a matrix of with the right singular vectors of mat in the columns. When cells are in the columns, these are the cell embeddings. (from SVD output)  
**eigsum** sum of the eigenvalues for calculating percent variance explained

For SingleCellExperiment input, returns the SCE with embeddings in the reducedDim slot 'corralm'

For a list of [SingleCellExperiments](#), returns a list of the SCEs with the embeddings in the respective reducedDim slot 'corralm'

.

## Examples

```

listofmats <- list(matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 25),
                  matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 25))
result <- corralm_matlist(listofmats)
library(DuoClustering2018)
library(SingleCellExperiment)
sce <- sce_full_Zhengmix4eq()[1:100,sample(1:3500,100,replace = FALSE)]
colData(sce)$Method <- matrix(sample(c('Method1','Method2'),100,replace = TRUE))
result <- corralm_sce(sce, splitby = 'Method')

listofmats <- list(matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 20),
                  matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 20))
corralm(listofmats)

library(DuoClustering2018)
library(SingleCellExperiment)
sce <- sce_full_Zhengmix4eq()[seq(1,100,1),sample(seq(1,3500,1),100,replace = FALSE)]
colData(sce)$Method <- matrix(sample(c('Method1','Method2'),100,replace = TRUE))
result <- corralm(sce, splitby = 'Method')

# default print method for corralm objects

```

---

corral\_mat

*corral: Correspondence analysis on a single matrix*

---

## Description

corral can be used for dimension reduction to find a set of low-dimensional embeddings for a count matrix.

corral is a wrapper for [corral\\_mat](#) and [corral\\_sce](#), and can be called on any of the acceptable input types.

**Usage**

```

corral_mat(
  inp,
  method = c("irl", "svd"),
  ncomp = 30,
  row.w = NULL,
  col.w = NULL,
  rtype = c("standardized", "indexed", "hellinger", "freemantukey", "pearson"),
  vst_mth = c("none", "sqrt", "freemantukey", "anscombe"),
  ...
)

corral_sce(
  inp,
  method = c("irl", "svd"),
  ncomp = 30,
  whichmat = "counts",
  fullout = FALSE,
  subset_row = NULL,
  ...
)

corral(inp, ...)

## S3 method for class 'corral'
print(x, ...)

```

**Arguments**

<code>inp</code>	matrix (any type), <code>SingleCellExperiment</code> , or <code>SummarizedExperiment</code> . If using <code>SingleCellExperiment</code> or <code>SummarizedExperiment</code> , then include the <code>whichmat</code> argument to specify which slot to use (defaults to counts).
<code>method</code>	character, the algorithm to be used for svd. Default is <code>irl</code> . Currently supports <code>'irl'</code> for <code>irlba::irlba</code> or <code>'svd'</code> for <code>stats::svd</code>
<code>ncomp</code>	numeric, number of components; Default is 30
<code>row.w</code>	numeric vector; the row weights to use in chi-squared scaling. Defaults to <code>'NULL'</code> , in which case row weights are computed from the input matrix.
<code>col.w</code>	numeric vector; the column weights to use in chi-squared scaling. For instance, size factors could be given here. Defaults to <code>'NULL'</code> , in which case column weights are computed from the input matrix.
<code>rtype</code>	character indicating what type of residual should be computed; options are <code>"indexed"</code> , <code>"standardized"</code> (or <code>"pearson"</code> is equivalent), <code>"freemantukey"</code> , and <code>"hellinger"</code> ; defaults to <code>"standardized"</code> for <code>corral</code> and <code>"indexed"</code> for <code>corralm</code> . <code>"indexed"</code> , <code>"standardized"</code> , and <code>"freemantukey"</code> compute the respective chi-squared residuals and are appropriate for count data. The <code>"hellinger"</code> option is appropriate for continuous data.

vst_mth	character indicating whether a variance-stabilizing transform should be applied prior to calculating chi-squared residuals; defaults to "none"
...	(additional arguments for methods)
whichmat	character; defaults to counts, can also use logcounts or normcounts if stored in the sce object
fullout	boolean; whether the function will return the full corral output as a list, or a SingleCellExperiment; defaults to SingleCellExperiment (FALSE). To get back the <code>corral_mat</code> -style output, set this to TRUE.
subset_row	numeric, character, or boolean vector; the rows to include in corral, as indices (numeric), rownames (character), or with booleans (same length as the number of rows in the matrix). If this parameter is NULL, then all rows will be used.
x	(print method) corral object; the list output from <code>corral_mat</code>

### Value

When run on a matrix, a list with the correspondence analysis matrix decomposition result:

`d` a vector of the diagonal singular values of the input mat (from SVD output)

`u` a matrix of with the left singular vectors of mat in the columns (from SVD output)

`v` a matrix of with the right singular vectors of mat in the columns. When cells are in the columns, these are the cell embeddings. (from SVD output)

`eigsum` sum of the eigenvalues for calculating percent variance explained

`SCu` and `SCv` standard coordinates, left and right, respectively

`PCu` and `PCv` principal coordinates, left and right, respectively

When run on a `SingleCellExperiment`, returns a SCE with the embeddings (`PCv` from the full corral output) in the `reducedDim` slot `corral` (default). Also can return the same output as `corral_mat` when `fullout` is set to TRUE.

For matrix and `SummarizedExperiment` input, returns list with the correspondence analysis matrix decomposition result (`u,v,d` are the raw svd output; `SCu` and `SCv` are the standard coordinates; `PCu` and `PCv` are the principal coordinates)

For `SummarizedExperiment` input, returns the same as for a matrix.

.

### Examples

```
mat <- matrix(sample(0:10, 5000, replace=TRUE), ncol=50)
result <- corral_mat(mat)
result <- corral_mat(mat, method = 'irl', ncomp = 5)
```

```
library(DuoClustering2018)
sce <- sce_full_Zhengmix4eq()[1:100,1:100]
result_1 <- corral_sce(sce)
result_2 <- corral_sce(sce, method = 'svd')
result_3 <- corral_sce(sce, method = 'irl', ncomp = 30, whichmat = 'logcounts')
```

```

library(DuoClustering2018)
sce <- sce_full_Zhengmix4eq()[1:100,1:100]
corral_sce <- corral(sce,whichmat = 'counts')

mat <- matrix(sample(0:10, 500, replace=TRUE), ncol=25)
corral_mat <- corral(mat, ncomp=5)

mat <- matrix(sample(1:100, 10000, replace = TRUE), ncol = 100)
corral(mat)

```

---

corral_preproc	<i>Preprocess a matrix for SVD to perform Correspondence Analysis (CA)</i>
----------------	--

---

### Description

This function performs the row and column scaling pre-processing operations, prior to SVD, for the corral methods. See [corral](#) for single matrix correspondence analysis and [corralm](#) for multi-matrix correspondence analysis.

### Usage

```

corral_preproc(
  inp,
  rtype = c("standardized", "indexed", "hellinger", "freemantukey", "pearson"),
  vst_mth = c("none", "sqrt", "freemantukey", "anscombe"),
  powdef_alpha = NULL,
  row.w = NULL,
  col.w = NULL,
  smooth = FALSE,
  ...
)

```

### Arguments

inp	matrix, numeric, counts or logcounts; can be sparse Matrix or matrix
rtype	character indicating what type of residual should be computed; options are "indexed", "standardized" (or "pearson" is equivalent), "freemantukey", and "hellinger"; defaults to "standardized" for <a href="#">corral</a> and "indexed" for <a href="#">corralm</a> . "indexed", "standardized", and "freemantukey" compute the respective chi-squared residuals and are appropriate for count data. The "hellinger" option is appropriate for continuous data.
vst_mth	character indicating whether a variance-stabilizing transform should be applied prior to calculating chi-squared residuals; defaults to "none"
powdef_alpha	numeric for the power that should be applied if using power deflation. Must be in (0,1), and if provided a number outside this range, will be ignored. Defaults to 'NULL' which does not perform this step.

row.w	numeric vector; Default is NULL, to compute row.w based on inp. Use this parameter to replace computed row weights with custom row weights
col.w	numeric vector; Default is NULL, to compute col.w based on inp. Use this parameter to replace computed column weights with custom column weights
smooth	logical; Whether or not to perform the additional smoothing step with 'trim_matdist'. Default is FALSE. Incompatible with 'powdef_alpha', so that parameter takes precedence over this one.
...	(additional arguments for methods)

**Value**

matrix, processed for input to compsvd to finish CA routine

**Examples**

```
mat <- matrix(sample(0:10, 500, replace=TRUE), ncol=25)
mat_corral <- corral_preproc(mat)
corral_output <- compsvd(mat_corral, ncomp = 5)
```

---

earthmover\_dist      *Earthmover distance (and general Wasserstein distance)*

---

**Description**

i.e., wasserstein distance with L1 ( $p\_param = 1$ ); can also use other penalties  $> 1$  (Not technically earthmover distance if using other  $p\_param$  values)

**Usage**

```
earthmover_dist(batch1, batch2, whichdim = 1, numbins = 100, p_param = 1)
```

**Arguments**

batch1	matrix; subset of observations from an embedding corresponding to some attribute (e.g., batch or phenotype)
batch2	matrix; subset of observations from an embedding corresponding to some attribute (e.g., batch or phenotype)
whichdim	int; which dimension (i.e., column) from the embeddings is used. defaults on first
numbins	int; number of bins for the probability discretization (defaults to 100)
p_param	int; penalty parameter for general Wasserstein distance. Defaults to 1, which corresponds to earthmover.

**Value**

num; the distance

**Examples**

```
# To compare distributions of reduced dimension values to assess similarity,
# e.g. as a metric for batch integration
embedding <- matrix(sample(x = seq(0,10,.1),1000, replace = TRUE),ncol = 5)
batch <- matrix(sample(c(1,2),200, replace = TRUE))
earthmover_dist(embedding[which(batch == 1),],embedding[which(batch == 2),])
```

---

get\_pct\_var\_exp\_svd     *Compute percent of variance explained*

---

**Description**

Compute percent of variance explained

**Usage**

```
get_pct_var_exp_svd(thissvd, preproc_mat = thissvd$d)
```

**Arguments**

thissvd	list outputted from an svd function (svd, irlba; can also take output from <a href="#">corral_mat</a> and <a href="#">corralm_matlist</a> )
preproc_mat	matrix of pre-processed values (optional) - important to include if the svd is only partial as this is used to compute the sum of eigenvalues

**Value**

vector of percent variance explained values, indexed by PC

**Examples**

```
mat <- matrix(sample(seq(0,20,1),100,replace = TRUE),nrow = 10)
my_svd <- svd(mat)
get_pct_var_exp_svd(my_svd) # this works if my_svd is a full svd
my_irl <- irlba::irlba(mat,nv = 2)
get_pct_var_exp_svd(my_irl, preproc_mat = mat) # ... otherwise use this
```

---

 get\_weights

*Get weights*


---

**Description**

Computes row weights and column weights

**Usage**

```
get_weights(inp_mat)
```

**Arguments**

inp\_mat            matrix for which weights should be calculated (sparse or full)

**Value**

list of 2 elements: 'row.w' and 'col.w' contain the row and column weights respectively

**Examples**

```
mat <- matrix(sample(seq(0,20,1),100,replace = TRUE),nrow = 10)
ws <- get_weights(mat)
```

---

 list2mat

*List to Matrix*


---

**Description**

List to Matrix

**Usage**

```
list2mat(matlist, direction = c("c", "r")[1])
```

**Arguments**

matlist            list of matrices to concatenate  
 direction        character, r or c, to indicate whether should be row-wise (i.e., rbind to match on columns) or column-wise (i.e., cbind to match on rows). Defaults to columnwise (matching on rows) to match convention of SingleCellExperiments

**Value**

matrix

**Examples**

```
listofmats <- list(matrix(sample(seq(0,20,1),100,replace = TRUE),nrow = 10),
                  matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 10))
newmat <- list2mat(listofmats) # to "cbind" them
listofmats_t <- lapply(listofmats,t)
newmat_t <- list2mat(listofmats_t, 'r') # to "rbind" them
```

---

na2zero	<i>Set na to 0</i>
---------	--------------------

---

**Description**

Set na to 0

**Usage**

```
na2zero(x)
```

**Arguments**

x                    matrix of values for which na values should be changed to 0

**Value**

matrix, where na values are set to 0

**Examples**

```
x <- matrix(sample(0:10, 5000, replace = TRUE), ncol = 25)
x[sample(1:5000, 10)] <- NA

na2zero(x)
```

---

obs2probs	<i>Observations -&gt; discrete probabilities</i>
-----------	--

---

**Description**

```
usage: embedding <- matrix(sample(x = seq(0,10,.1),200,replace = TRUE)) disc_probs <- obs2probs(embedding)
```

**Usage**

```
obs2probs(obs, numbins = 100, startbin = min(obs), endbin = max(obs) + 1e-05)
```

**Arguments**

obs	vector of numeric, with the observations
numbins	int, the number of evenly sized bins to discretize the observations to
startbin	numeric, the starting value for the smallest bin. Defaults to taking the minimum of obs
endbin	numeric, the ending value for the largest bin. Defaults to taking the maximum of obs (plus a tiny decimal to ensure full range of obs is captured)

**Value**

dataframe, results has rows corresponding to each bin with columns for probability ('prob'), cumulative frequency ('cumfreq'), and frequency ('freq') of observations falling into that bin. The 'bins' column indicates the end of the bin (start is the preceding column)

---

pairwise_rv	<i>Pairwise rv coefficient</i>
-------------	--------------------------------

---

**Description**

Pairwise rv coefficient

**Usage**

```
pairwise_rv(matlist)
```

**Arguments**

matlist	list of matrices (or matrix-like; see rv function) for which to compute pairwise RV coefficients
---------	--

**Value**

matrix of the pairwise coefficients

**Examples**

```
a <- matrix(sample(1:10,100,TRUE), nrow = 10)
b <- matrix(sample(1:10,50,TRUE), nrow = 5)
c <- matrix(sample(1:10,20,TRUE), nrow = 2)

matlist <- list(a,b,c)
pairwise_rv(matlist)
pairwise_rv(lapply(matlist, t))
```

---

plot_embedding	<i>Plot selected PCs from an embedding</i>
----------------	--

---

### Description

Plot selected PCs from an embedding

### Usage

```
plot_embedding(
  embedding,
  xpc = 1,
  ypc = xpc + 1,
  plot_title = paste0("Dim", xpc, " by Dim", ypc),
  color_vec = NULL,
  color_title = NULL,
  ellipse_vec = NULL,
  facet_vec = NULL,
  psize = 0.8,
  saveplot = FALSE,
  plotfn = paste(plot_title, xpc, sep = "_"),
  showplot = TRUE,
  returngg = FALSE,
  color_pal_vec = NULL,
  dimname = "Dim"
)
```

### Arguments

embedding	matrix or other tabular format where columns correspond to PCs and rows correspond to cells (entries). corral and corralm objects are also accepted.
xpc	int; which PC to put on the x-axis (defaults to 1)
ypc	int; which PC to put on the y-axis (defaults to the one after xpc)
plot_title	char; title of plot (defaults to titling based on xpc and ypc)
color_vec	vector; length should correspond to the number of rows in embedding, and each element of the vector classifies that cell (entry) in the embedding to that particular class, which will be colored the same. (e.g., this could be indicating which batch each cell is from)
color_title	char; what attribute the colors represent
ellipse_vec	vector; length should correspond to the number of rows in embedding, and each element of the vector classifies that cell (entry) in the embedding to that particular class, and elements of the same class will be circled in an ellipse. (e.g., this could be indicating the cell type or cell line; works best for attributes intended to be compact)

facet_vec	vector; length should correspond to the number of rows in embedding, and each element of the vector classifies that cell (entry) in the embedding to that particular class. Plot will be faceted by this attribute.
ptsize	numeric; the size of the points as passed to <code>geom_point()</code> . Defaults to 0.8.
saveplot	boolean; whether or not to save the plot, defaults FALSE
plotfn	char; what the filename is to be called. (defaults to making a name based on <code>plot_title</code> and <code>xpc</code> )
showplot	boolean; whether or not to show the plot, defaults TRUE
returngg	boolean; whether or not to return a <code>ggplot2</code> object, defaults FALSE
color_pal_vec	char; hex codes for the color palette to be used. Default is to use the <code>ggthemes</code> few for plots with less than 9 colors, and to use "stretch" pals polychrome if more colors are needed.
dimname	char; the name of the dimensions. defaults to "Dim"

**Value**

default none; options to display plot (`showplot`), save plot (`saveplot`), and/or return `ggplot2` object (`returngg`)

**Examples**

```
listofmats <- list(matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 20),
                  matrix(sample(seq(0,20,1),1000,replace = TRUE),nrow = 20))
corralm_obj <- corralm(listofmats, ncomp = 5)
embed_mat <- corralm_obj$v
cell_type_vec <- sample(c('type1','type2','type3'),100,replace = TRUE)
plot_embedding(embedding = embed_mat,
              xpc = 1,
              plot_title = 'corralm plot',
              color_vec = cell_type_vec,
              color_title = 'cell type',
              saveplot = FALSE)

# or, call directly on the corralm object
plot_embedding(corralm_obj)
```

---

plot\_embedding\_sce      *Plot selected PCs from an embedding saved in a SingleCellExperiment object*

---

**Description**

Plot selected PCs from an embedding saved in a SingleCellExperiment object

**Usage**

```
plot_embedding_sce(
  sce,
  which_embedding,
  color_attr = NULL,
  color_title = color_attr,
  ellipse_attr = NULL,
  facet_attr = NULL,
  ...
)
```

**Arguments**

sce	<a href="#">SingleCellExperiment</a> object; contains the embedding within the reducedDim slot
which_embedding	character; for the embedding to plot
color_attr	character; name of the attribute within colData to use for assigning colors (in lieu of color_vec in the <a href="#">plot_embedding</a> function)
color_title	character; title to use for colors legend, defaults to the same as color_attr
ellipse_attr	character; name of the attribute within colData to use for drawing ellipse(s) (in lieu of ellipse_vec in the <a href="#">plot_embedding</a> function)
facet_attr	character; name of the attribute within colData to use for faceting (in lieu of facet_vec in the <a href="#">plot_embedding</a> function)
...	additional optional arguments - see <a href="#">plot_embedding</a> function for details on other potential arguments: xpc, ypc, plot_title, color_title (if title is different from color_attr), ptsize, saveplot, plotfn, showplot, returngg, color_pal_vec, dimname

**Value**

default none; options to display plot (showplot), save plot (saveplot), and/or return [ggplot2](#) object (returngg)

**Examples**

```
library(DuoClustering2018)
library(SingleCellExperiment)
sce <- sce_full_Zhengmix4eq()[1:100, sample(1:3500, 100, replace = FALSE)]
colData(sce)$Method <- matrix(sample(c('Method1', 'Method2'), 100, replace = TRUE))
sce <- corralm(sce, splitby = 'Method')

# to plot and show only
plot_embedding_sce(sce = sce,
  which_embedding = 'corralm',
  xpc = 1,
  plot_title = 'corralm: PC1 by PC2',
  color_attr = "Method",
```

```

        ellipse_attr = 'phenoid',
        saveplot = FALSE)

# to return ggplot2 object and display, but not save
corralm_ggplot <- plot_embedding_sce(sce = sce,
                                   which_embedding = 'corralm',
                                   xpc = 1,
                                   plot_title = 'corralm: PC1 by PC2',
                                   color_attr = 'Method',
                                   ellipse_attr = 'phenoid',
                                   returngg = TRUE,
                                   saveplot = FALSE)

```

---

rv	<i>rv coefficient</i>
----	-----------------------

---

### Description

rv coefficient

### Usage

```
rv(mat1, mat2)
```

### Arguments

mat1	matrix (or matrix-like, e.g., df); either columns or rows should be matched with mat2
mat2	matrix (or matrix-like, e.g., df); either columns or rows should be matched with mat1

### Value

numeric; RV coefficient between the matched matrices

### Examples

```

a <- matrix(sample(1:10,100, TRUE), nrow = 10)
b <- matrix(sample(1:10,50, TRUE), nrow = 5)

rv(a, b) # matched by columns
rv(t(a), t(b)) # matched by rows

```

---

scal_var	<i>Generate a scaled variance plot for an integrative embedding</i>
----------	---

---

## Description

Generate a scaled variance plot for an integrative embedding

## Usage

```
scal_var(  
  inp,  
  batchvec = NULL,  
  pcs = seq(3),  
  returngg = FALSE,  
  showplot = TRUE,  
  plot_subtitle = NULL  
)
```

## Arguments

inp	corralm object or matrix; embedding to compute scaled variances
batchvec	vector; batch labels (can be numeric or char). Defaults to 'NULL', which is appropriate for using a corralm object. If using an embedding matrix for inp, then this argument must be given and length must correspond to number of rows in 'inp'.
pcs	numeric; vector of which PCs should be shown. Defaults to 1:3
returngg	boolean; whether or not to return a <a href="#">ggplot2</a> object, defaults FALSE
showplot	boolean; whether or not to show the plot, defaults TRUE
plot_subtitle	string; the text that should show in the subtitle for the plot. defaults to NULL

## Value

N/A or a ggplot object

## Examples

```
dat <- matrix(rnorm(10000), ncol = 50)  
bv <- rep(seq(4),c(10,30,60,100))  
scal_var(dat,bv, pcs = seq(4))
```

---

scal_var_mat	<i>Generate a matrix of the scaled variance values</i>
--------------	--

---

**Description**

Generate a matrix of the scaled variance values

**Usage**

```
scal_var_mat(inp, batchvec = NULL)
```

**Arguments**

inp	corralm object or matrix; embedding to compute scaled variances
batchvec	vector; batch labels (can be numeric or char). Defaults to 'NULL', which is appropriate for using a corralm object. If using an embedding matrix for inp, then this argument must be given and length must correspond to number of rows in 'inp'.

**Value**

matrix of the scaled variance values by PC (batches in rows; PCs in columns)

**Examples**

```
dat <- matrix(rnorm(5000), ncol = 50)
bv <- rep(seq(3), c(10, 30, 60))
scal_var_mat(dat, bv)
```

---

sce2matlist	<i>SingleCellExperiment to list of matrices</i>
-------------	---

---

**Description**

SingleCellExperiment to list of matrices

**Usage**

```
sce2matlist(sce, splitby, to_include = NULL, whichmat = "counts")
```

**Arguments**

sce	SingleCellExperiment that is to be separated into list of count matrices
splitby	character; name of the attribute from colData that should be used to separate the SCE
to_include	(optional) character vector; determines which values from the "splitby" column will be included in the outputted matlist. NULL is the default, and will result in selecting all elements
whichmat	character; defaults to counts, can also use logcounts or normcounts if stored in the sce object

**Value**

list of matrices

**Examples**

```
library(DuoClustering2018)
sce <- sce_full_Zhengmix4eq()
matlist <- sce2matlist(sce = sce, splitby = 'phenoid', whichmat = 'logcounts')
```

---

trim_matdist	<i>Trim extreme values in a pre-processed matrix</i>
--------------	--

---

**Description**

Smooths the extreme values in a chi-square-transformed matrix to lessen the influence of "rare objects."

**Usage**

```
trim_matdist(mat, pct_trim = 0.01)
```

**Arguments**

mat	matrix; should be pre-processed/normalized to some sort of approximately normally distributed statistic (e.g., chi-squared transformation with 'corral_preproc' or Z-score normalization)
pct_trim	numeric; the percent of observations to smooth. Defaults to 'pct_trim' = .01, which corresponds to smoothing all observations to be between the .5 percentile and 99.5 percentile range of the input matrix

**Details**

(Usually not called directly; can be included by using the 'smooth' argument in the 'corral', 'corralm', and 'corral\_preproc' functions)

**Value**

smoothed matrix

**Examples**

```
count_mat <- matrix(rpois(10000, 300)*rbinom(10000,1,.1), ncol = 100)
smoothed_preproc_mat <- corral_preproc(count_mat, smooth = TRUE)
```

---

var_stabilize	<i>Apply a variance stabilizing transformation</i>
---------------	--

---

**Description**

Prior to running CA, there is an option to apply a variance stabilizing transformation. This function can be called explicitly or used with the 'vst\_mth' argument in corral and corral\_preproc.

**Usage**

```
var_stabilize(inp, transform = c("sqrt", "freemantuke", "anscombe"))
```

**Arguments**

inp	matrix, numeric, counts or logcounts; can be sparse Matrix or matrix
transform	character indicating which method should be applied. Defaults to the square root transform ("sqrt"). Other options include "freemantuke" and "anscombe".

**Value**

variance-stabilized matrix; sparse if possible

**Examples**

```
x <- as.matrix(rpois(100, lambda = 50), ncol = 10)
vst_x <- var_stabilize(x)
```

# Index

- \* **internal**
  - obs2probs, 15
- add\_embeddings2scelist, 2
- all\_are, 3
- biplot\_corral, 4
- compsvd, 5
- corral, 7, 9, 11
- corral (corral\_mat), 8
- corral\_mat, 8, 8, 10, 13
- corral\_preproc, 5, 11
- corral\_sce, 8
- corral\_sce (corral\_mat), 8
- corralm, 3, 7, 9, 11
- corralm (corralm\_matlist), 6
- corralm\_matlist, 6, 7, 13
- corralm\_sce, 7
- corralm\_sce (corralm\_matlist), 6
- earthmover\_dist, 12
- get\_pct\_var\_exp\_svd, 13
- get\_weights, 14
- ggplot2, 18, 19, 21
- list2mat, 14
- na2zero, 15
- obs2probs, 15
- pairwise\_rv, 16
- plot\_embedding, 17, 19
- plot\_embedding\_sce, 18
- print.corral (corral\_mat), 8
- print.corralm (corralm\_matlist), 6
- rv, 20
- scal\_var, 21
- scal\_var\_mat, 22
- sce2matlist, 7, 22
- SingleCellExperiment, 8, 10, 19
- trim\_matdist, 23
- var\_stabilize, 24