Package 'ADM'

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Description ADM provides a suite of advanced distance measurement methods for high-

Title Advanced Distance Measurement for High-Dimensional Data Analysis

dimensional data analysis. The package implements various dimensionality reduction and visualization techniques, including UMAP, t-SNE, and PCA, along with tools for evaluating the performance of these methods. Additionally, ADM includes functions for calculating intergroup similarities and custom ggplot2 themes for data visualization.

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adm

Multidimensional Embedding Visualization (ADM)

Description

Multidimensional Embedding Visualization (ADM)

Usage

```
adm(
    e,
    k.dim = NULL,
    dist.power = 0.5,
    conn.prop = 0.02,
    raw.d.pwr = 0.5,
    diffu.steps = NA,
    diffu.factor = 3.5,
    distr.template = "gamma",
    gamma.shape = 3,
    gamma.rate = 3,
    scale.dist = TRUE,
    symmetrize = "mean",
    dist.quantile = 0.25
)
```

Arguments

e A list of matrices, each representing a dimension reduction result.	
Number of dimensions to use. If NULL, uses the number of columns in the matrix of e.	
dist.power	Power to raise distances to when merging. Default is 0.5.
conn.prop	Proportion of connections to consider. Default is 0.02.
raw.d.pwr	Power for raw distance calculation. Default is 0.5.
diffu.steps	Steps for diffusion. If NA, calculated based on graph properties.
diffu.factor	Factor to multiply diffusion steps by. Default is 3.5.

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distr.template	Distribution template to use. Options are "none", "combine", "gamma", or "parametric". Default is "gamma".	
gamma.shape	Shape parameter for gamma distribution. Default is 3.	
gamma.rate	Rate parameter for gamma distribution. Default is 3.	
scale.dist	Whether to scale distances. Default is TRUE.	
symmetrize	Method to symmetrize distance matrix. Options are "mean", "min", or "max". Default is "mean".	
dist.quantile	Quantile to use for distance calculations. Default is 0.25.	

Value

A list containing the final merged distance matrix.

cal_ari_nmi	Calculate Adjusted Rand Index (ARI) and Normalized Mutual Information (NMI)

Description

Calculate Adjusted Rand Index (ARI) and Normalized Mutual Information (NMI)

Usage

```
cal_ari_nmi(lowDim_data, k, method_name, seed, info = NULL)
```

Arguments

lowDim_data A matrix or data frame of low-dimensional data.

k The number of clusters for k-means clustering.

method_name A string naming the method being evaluated.

seed An integer seed for reproducibility.

info A vector of true class labels. If NULL, it will use the global 'info' variable.

Details

This function performs k-means clustering on the input data and calculates the Adjusted Rand Index and Normalized Mutual Information between the clustering result and a pre-existing classification.

Value

A data frame with ARI and NMI scores.

Examples

```
data <- matrix(rnorm(200), 100, 2)
info <- sample(1:3, 100, replace = TRUE) # Simulated cluster information
result <- cal_ari_nmi(data, k = 3, method_name = "Example Method", seed = 123, info = info)</pre>
```

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cal_cci	Calculate Cluster Conservation Index (CCI)

Description

This function calculates the Cluster Conservation Index (CCI) for both the ensemble and ADM methods.

Usage

```
cal_cci(ensemble.out, mev.out, info)
```

Arguments

ensemble.out	A list containing the output from the ensemble method, including the ensemble distance matrix.
mev.out	A list containing the output from the ADM method, including the diffusion distance matrix.
info	A vector or factor containing cluster information for each data point.

Details

The function calculates CCI for different numbers of neighbors (1, 2, 5, 10, 20) and for different representations of the data (raw, UMAP, PCA).

Value

A list containing two elements:

cci_viz	A matrix with CCI values for the ensemble method
cci_adm	A matrix with CCI values for the ADM method

candidate.visual	Perform Multiple Dimensionality	Reduction Methods
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Description

This function applies various dimensionality reduction methods to the input data. It supports methods such as PCA, MDS, iMDS, Sammon mapping, LLE, HLLE, Isomap, kPCA, LEIM, UMAP, t-SNE, PHATE, and KEF.

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Usage

```
candidate.visual(
  data,
  dim = 2,
methods = c("PCA", "MDS", "iMDS", "Sammon", "HLLE", "Isomap", "kPCA", "LEIM", "UMAP",
        "tSNE", "PHATE", "KEF"),
  kpca.sigma = c(0.001, 0.002),
  umap.k = c(30, 50),
  tsne.perplexity = c(30, 50),
  phate.k = c(30, 50),
  cal_dist = TRUE
)
```

Arguments

A numeric matrix or data frame where rows are observations and columns are features.

dim Integer. The number of dimensions for the reduced space. Default is 2.

methods Character vector. Methods to be applied. Default includes all supported methods.

kpca.sigma Numeric vector. Sigma values for kPCA. Default is c(0.001, 0.002).

umap.k Integer vector. Number of neighbors for UMAP. Default is c(30, 50).

tsne.perplexity Numeric vector. Perplexity values for t-SNE. Default is c(30, 50).

 $\label{eq:phate.k} \textbf{Integer vector. Number of nearest neighbors for PHATE. Default is $c(30,50)$.}$

cal_dist Logical. Whether to calculate distance matrix. Default is TRUE.

Details

The function applies each specified method to the input data. For some methods (kPCA, UMAP, t-SNE, PHATE), multiple parameter settings are tried, resulting in multiple outputs for these methods.

Value

A list containing two elements:

embed.list A list of matrices, each representing the reduced data for a method method_name A character vector of method names corresponding to embed.list

Note

- This function requires several packages including rARPACK, MASS, dimRed, umap, and phateR.
- The function prints progress messages to the console. Some methods may fail if the required packages are not installed.

Examples

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```
umap.k = c(15, 30))
pca_result <- result$embed.list[[1]]
method_names <- result$method_name
## End(Not run)</pre>
```

dataloader

Load and Prepare Dataset for Analysis

Description

Load and Prepare Dataset for Analysis

Usage

```
dataloader(dataset, base_dir = ".")
```

Arguments

dataset A string specifying the dataset to load. Supported datasets are: "Oihane", "Gutier-

rez", "Spleen", "Quake", "pbmc", "metabolism", "kidney", "mir", "gene".

base_dir A string specifying the base directory for data files. Default is ".".

Details

This function loads and prepares various datasets for analysis. Each dataset is processed differently based on its structure and content. The function sets the working directory to a specific path for each dataset before loading.

Value

A list containing:

dat The main data matrix or data frame

cell.type A factor of cell types

info A factor of additional information (often identical to cell.type)

Note

- The function assumes a specific directory structure and file naming convention. - Some datasets require additional files to be present in the specified directories. - For "mir" and "gene" datasets, extensive preprocessing is performed.

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Examples

```
## Not run:
# Load the Oihane dataset
oihane_data <- dataloader("Oihane")
# Load the PBMC dataset
pbmc_data <- dataloader("pbmc")
# Load the miRNA dataset
mir_data <- dataloader("mir")
## End(Not run)</pre>
```

dist.grp

Calculate distance-based group statistics

Description

This function calculates distance-based group statistics for a given distance matrix and grouping.

Usage

```
dist.grp(distmat, grp, k = 1:20)
```

Arguments

distmat A distance matrix.

grp A vector of group labels for each data point.

k Vector of integers; the number of neighbors to consider.

Value

A matrix with two columns: the k values and corresponding distance-based group statistics.

get_mapping Get label mapping for a dataset

Description

This function returns the appropriate label mapping based on the provided dataset name. It currently supports four datasets: "Oihane", "mir", "gene", and "Quake".

Usage

```
get_mapping(dataset)
get_mapping(dataset)
```

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Arguments

dataset

A string specifying the name of the dataset. Possible values are "Oihane", "mir", "gene", or "Quake".

Value

A named vector where names are the original labels and values are the corresponding abbreviations or simplified labels. Returns NULL if the provided dataset name is not recognized.

A named vector where names are the original labels and values are the corresponding abbreviations or simplified labels. Returns NULL if the provided dataset name is not recognized.

Examples

```
get_mapping("Oihane")
get_mapping("Quake")
get_mapping("Oihane")
get_mapping("Quake")
```

move.outlier

Move Outliers Function

Description

Move Outliers Function

Usage

```
move.outlier(x, d = NULL, fraction = 0.01)
```

Arguments

x A numeric vector or matrix.

d The distance threshold for outlier detection. If NULL, it's calculated from the

data.

fraction The fraction of points to consider as outliers.

Value

A numeric vector or matrix with outliers moved.

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Plot Comparison of CCI Results

Description

This function processes and visualizes the results of Cell-Cell Interaction (CCI) analyses for different methods and data transformations.

Usage

```
plot_cci_results(cci_list, dataset)
```

Arguments

cci_list A list containing two matrices: one for visualization method and one for diffu-

sion method.

dataset A string specifying the name of the dataset (used for plot title).

Details

The function processes CCI matrices for visualization and diffusion methods, combines the data, and creates a boxplot comparing the results across different data types (raw, UMAP, PCA) and methods (metaspec, ADM).

Value

A ggplot object representing a boxplot of CCI results.

Examples

plot_legend

Create a scatter plot with a custom legend

Description

This function generates a scatter plot from UMAP coordinates with a customized legend.

Usage

```
plot_legend(
   umap_viz,
   method_name,
   info,
   dataset,
   color_list,
   label_mapping = NULL
)
```

Arguments

umap_viz A matrix or data frame with UMAP coordinates (at least two columns).
 method_name A string specifying the visualization method name (used for plot title).
 info A vector of original cluster labels for each data point.
 dataset A string specifying the dataset name (not used in the current implementation).
 color_list A named vector of colors for each cluster.
 label_mapping An optional named vector for mapping original labels to simplified ones.

Details

The function creates a scatter plot of UMAP coordinates, colors points by cluster, and applies a custom theme with a specially formatted legend. It uses the 'simplify_labels' function to potentially simplify cluster labels.

Value

A ggplot object representing the scatter plot with a customized legend.

Examples

```
umap_coords <- matrix(rnorm(200), ncol = 2)
cluster_info <- rep(c("Type A", "Type B"), each = 50)
color_list <- c("Type A" = "red", "Type B" = "blue")
label_map <- c("Type A" = "A", "Type B" = "B")
p <- plot_legend(umap_coords, "UMAP", cluster_info, "Example", color_list, label_map)
if (interactive()) {
   print(p)
}</pre>
```

Description

Process and Visualize Meta-Methods

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Usage

```
process_and_visualize_meta_methods(
  mev.out,
  ensemble.out = NULL,
  info,
  k,
  color_list,
  seed = 2024
)
```

Arguments

mev.out Output from MEV method (assumed to contain \$diffu.dist).

ensemble.out Output from an ensemble method (assumed to contain \$ensemble.dist.mat). Can

be NULL.

info A vector of true class labels.

k The number of clusters for k-means clustering.

color_list A list of colors for visualization.

seed An integer seed for reproducibility. Default is 2024.

Details

This function processes and visualizes results from two meta-methods: meta-spec and ADM. If ensemble out is NULL, only ADM method is executed. It performs UMAP dimensionality reduction, k-means clustering, and calculates various metrics.

Value

A list containing ARI and NMI scores, silhouette widths, and UMAP coordinates for meta-spec and/or ADM methods.

Note

This function requires the following packages: umap, mclust, aricode, cluster

Examples

```
# This function requires specific input structures and external functions
# An example cannot be easily provided without those dependencies
```

simplify_labels

Simplify labels based on a provided mapping

Description

This function simplifies a vector of labels using an optional mapping. If no mapping is provided, it returns the original labels.

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Usage

```
simplify_labels(info, mapping = NULL)
```

Arguments

info A vector of original labels to be simplified.

mapping An optional named vector where names are original labels and values are sim-

plified labels. Default is NULL.

Details

If a mapping is provided, the function replaces each label in info with its corresponding value in the mapping. For any labels not found in the mapping, the original label is retained.

Value

A vector of simplified labels. If no mapping is provided, returns the original labels. If mapping is provided, returns mapped labels where possible, falling back to original labels where no mapping exists.

Examples

```
# Without mapping
original_labels <- c("Type A", "Type B", "Type C")
simplify_labels(original_labels)
# Returns: c("Type A", "Type B", "Type C")

# With mapping
mapping <- c("Type A" = "A", "Type B" = "B")
simplify_labels(original_labels, mapping)
# Returns: c("A", "B", "Type C")</pre>
```

umap5

Perform UMAP and calculate distance-based group statistics

Description

This function applies UMAP to the input data and calculates distance-based group statistics.

Usage

```
umap5(x, info, do.plot = TRUE, n_components, k = c(1, 2, 5, 10, 20))
```

Arguments

x A distance matrix or object that can be coerced to one.

info A vector of group labels for each data point.

do.plot Logical; if TRUE, creates a pairs plot of the UMAP result.

n_components Integer; the number of dimensions for UMAP.

k Vector of integers; the number of neighbors to consider for distance calculations.

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Value

A vector of mean distance-based group statistics across multiple UMAP runs.

visualization_func Visualize data using ggplot2

Description

This function creates a scatter plot of the input data, optionally coloring points by group.

Usage

```
visualization_func(data, method_name, color_list = NULL, info = NULL)
```

Arguments

data A matrix or data frame with at least two columns for x and y coordinates.

method_name A string specifying the name of the visualization method (used for plot title).

color_list An optional named vector of colors for each group in 'info'.

info An optional vector of group labels for each data point.

Value

A ggplot object representing the scatter plot.

```
visualization_with_label
```

Create a labeled scatter plot from UMAP visualization

Description

This function generates a scatter plot from UMAP coordinates with labeled cluster centers.

Usage

```
visualization_with_label(umap_viz, method, info, dataset, color_list)
```

Arguments

umap_viz A matrix or data frame with UMAP coordinates (at least two columns).

Method A string specifying the visualization method name (used for plot title).

info A vector of cluster labels for each data point.

dataset A string specifying the dataset name (not used in the current implementation).

color_list A named vector of colors for each cluster.

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Details

The function creates a scatter plot of UMAP coordinates, colors points by cluster, calculates and labels cluster centers, and applies a custom theme. It uses ggrepel for non-overlapping text labels.

Value

A ggplot object representing the labeled scatter plot.

Examples

```
umap_coords <- matrix(rnorm(200), ncol = 2)
cluster_info <- rep(c("A", "B"), each = 50)
color_list <- c("A" = "red", "B" = "blue")
p <- visualization_with_label(umap_coords, "UMAP", cluster_info, "Example", color_list)
if (interactive()) {
   print(p)
}</pre>
```

visualize_ari_nmi

Visualize ARI and NMI Comparison with Bar Plot

Description

This function generates a bar plot to compare ARI and NMI metrics across methods.

Usage

```
visualize_ari_nmi(ARI_list, dataset)
```

Arguments

ARI_list A list containing one or two elements: ARI and NMI scores for ADM method,

and optionally for meta-spec method.

dataset A string naming the dataset being visualized.

Details

The function adapts to whether one (ADM only) or two (meta-spec and ADM) methods are present in the input. It creates a bar plot using ggplot2, with metrics on the x-axis and their values on the y-axis. Bars are grouped and colored by the 'group' variable. The plot uses a minimal theme and predefined colors.

Value

Invisibly returns a ggplot object visualizing ARI and NMI scores.

Examples

```
# Example with only ADM method
ARI_list_single <- list(data.frame(ARI = 0.7, NMI = 0.8))
visualize_ari_nmi(ARI_list_single, "Dataset A")

# Example with both methods
ARI_list_double <- list(
   data.frame(ARI = 0.8, NMI = 0.9),
   data.frame(ARI = 0.7, NMI = 0.8)
)
visualize_ari_nmi(ARI_list_double, "Dataset B")</pre>
```

visualize_individual_methods

Visualize Individual Methods

Description

This function creates scatter plots for multiple matrices using different methods.

Usage

```
visualize_individual_methods(
  matrices,
  method_names,
  info = NULL,
  color_list = NULL
)
```

Arguments

matrices A list of matrices, each representing a different method's output.

method_names A character vector specifying the names of the visualization methods.

An optional vector of group labels for each data point. If not provided, all points will be the same color.

color_list An optional named vector of colors for each group in 'info'. If not provided, a

default color palette will be used.

Value

A list of ggplot objects representing the scatter plots.

Examples

```
matrices <- list(matrix(rnorm(200), 100, 2), matrix(rnorm(200), 100, 2))
method_names <- c("Method1", "Method2")
# Example with default coloring (all points same color)
plots1 <- visualize_individual_methods(matrices, method_names)
# Example with group information
info <- sample(c("A", "B", "C"), 100, replace = TRUE)</pre>
```

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```
color_list <- c("A" = "red", "B" = "blue", "C" = "green")
plots2 <- visualize_individual_methods(matrices, method_names, info, color_list)</pre>
```

```
visualize_silhouette_width
```

Visualize Silhouette Width Comparison

Description

This function creates a boxplot to compare Silhouette Width values across different clusters and methods.

Usage

```
visualize_silhouette_width(ASW_list, info, data_name, label_mapping = NULL)
```

Arguments

ASW_list A list of vectors containing Silhouette Width values for each method.

info A vector of cluster labels for each data point.

data_name A string specifying the name of the dataset (used for plot title).

label_mapping An optional named vector for mapping original cluster labels to simplified ones.

Value

A ggplot object representing a boxplot of Silhouette Width values.

Examples

```
ASW_list <- list(Method1 = rnorm(100, 0.7, 0.1), Method2 = rnorm(100, 0.6, 0.1)) info <- sample(c("A", "B", "C"), 100, replace = TRUE) label_mapping <- c("A" = "Cluster 1", "B" = "Cluster 2", "C" = "Cluster 3") p <- visualize_silhouette_width(ASW_list, info, "Example Dataset", label_mapping) if (interactive()) { print(p) }
```

zp.quantile

Quantile Normalization Function

Description

Quantile Normalization Function

Usage

```
zp.quantile(x, y)
```

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Arguments

X	A numeric vector.	The reference	vector whose	quantiles w	ill be used	for nor-
	malization.					

y A numeric vector. The vector to be normalized based on the quantiles of 'x'.

Value

A numeric vector of the same length as 'y', containing the quantile-normalized values.

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