

Package ‘easybio’

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Title Comprehensive Single-Cell Annotation and Transcriptomic Analysis Toolkit

Version 1.2.3

Description Provides a comprehensive toolkit for single-cell annotation with the 'Cell-Marker2.0' database (see Xia Li, Peng Wang, Yun-peng Zhang (2023) <[doi:10.1093/nar/gkac947](https://doi.org/10.1093/nar/gkac947)>). Streamlines biological label assignment in single-cell RNA-seq data and facilitates transcriptomic analysis, including preparation of TCGA<<https://portal.gdc.cancer.gov/>> and GEO<<https://www.ncbi.nlm.nih.gov/geo/>> datasets, differential expression analysis and visualization of enrichment analysis results. Additional utility functions support various bioinformatics workflows. See Wei Cui (2024) <[doi:10.1101/2024.09.14.609619](https://doi.org/10.1101/2024.09.14.609619)> for more details.

URL <https://github.com/person-c/easybio>

BugReports <https://github.com/person-c/easybio/issues>

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Description

The `Artist` class offers a suite of methods designed to create a variety of plots using `ggplot2` for data exploration. Any methods prefixed with `plot_` or `test_` will log the command history along with their results, allowing you to review all outcomes later via the `get_all_results()` method. Notably, methods starting with `plot_` will check if the result of the preceding command is of the `htest` class. If so, it will incorporate the previous command and its p-value as the title and subtitle, respectively. This class encompasses methods for crafting dumbbell plots, bubble plots, divergence bar charts, lollipop plots, contour plots, scatter plots with ellipses, donut plots, and pie charts. Each method is tailored to map data to specific visual aesthetics and to apply additional customizations as needed.

Value

The R6 class [Artist](#).

Public fields

`data` Stores the dataset used for plotting.
`command` recode the command.
`result` record the plot.

Methods

Public methods:

- `Artist$new()`
- `Artist$get_all_result()`
- `Artist$test_wilcox()`
- `Artist$test_t()`
- `Artist$plot_scatter()`
- `Artist$plot_box()`
- `Artist$dumbbell()`
- `Artist$bubble()`
- `Artist$barchart_divergence()`
- `Artist$lollipop()`
- `Artist$contour()`
- `Artist$scatter_ellipses()`
- `Artist$donut()`
- `Artist$pie()`
- `Artist$clone()`

Method `new()`: Initializes the Artist class with an optional dataset.

Usage:

```
Artist$new(data = NULL)
```

Arguments:

`data` A data frame containing the dataset to be used for plotting. Default is `NULL`.

Returns: An instance of the Artist class.

Method `get_all_result()`: Get all history result

Usage:

```
Artist$get_all_result()
```

Returns: a `data.table` object

Method `test_wilcox()`: Conduct `wilcox.test`

Usage:

```
Artist$test_wilcox(formula, data = self$data, ...)
```

Arguments:

`formula` `wilcox.test()` formula arguments

`data` A data frame containing the data to be plotted. Default is `self$data`.

`...` Additional aesthetic mappings passed to `wilcox.test()`.

Returns: A `ggplot2` scatter plot.

Method `test_t()`: Conduct `wilcox.test`

Usage:

```
Artist$test_t(formula, data = self$data, ...)
```

Arguments:

`formula` `t.test()` formula arguments

`data` A data frame containing the data to be plotted. Default is `self$data`.

`...` Additional aesthetic mappings passed to `t.test()`.

Returns: A `ggplot2` scatter plot.

Method `plot_scatter()`: Creates a scatter plot.

Usage:

```
Artist$plot_scatter(
  data = self$data,
  fun = function(x) x,
  x,
  y,
  ...,
  add = private$is_htest()
)
```

Arguments:

`data` A data frame containing the data to be plotted. Default is `self$data`.

fun function to process the self\$data.
 x The column name for the x-axis.
 y The column name for the y-axis.
 ... Additional aesthetic mappings passed to aes().
 add whether to add the test result.
Returns: A ggplot2 scatter plot.

Method plot_box(): Creates a box plot.

Usage:

```
Artist$plot_box(
  data = self$data,
  fun = function(x) x,
  x,
  ...,
  add = private$is_htest()
)
```

Arguments:

data A data frame or tibble containing the data to be plotted. Default is self\$data.
 fun function to process the self\$data.
 x The column name for the x-axis.
 ... Additional aesthetic mappings passed to aes().
 add whether to add the test result.

Returns: A ggplot2 box plot.

Method dumbbell(): Create a dumbbell plot

This method generates a dumbbell plot using the provided data, mapping the specified columns to the x-axis, y-axis, and color aesthetic.

Usage:

```
Artist$dumbbell(data = self$data, x, y, col, ...)
```

Arguments:

data A data frame containing the data to be plotted.
 x The column in data to map to the x-axis.
 y The column in data to map to the y-axis.
 col The column in data to map to the color aesthetic.
 ... Additional aesthetic mappings or other arguments passed to ggplot.

Returns: A ggplot object representing the dumbbell plot.

Method bubble(): Create a bubble plot

This method generates a bubble plot where points are mapped to the x and y axes, with their size and color representing additional variables.

Usage:

```
Artist$bubble(data = self$data, x, y, size, col, ...)
```

Arguments:

data A data frame containing the data to be plotted.
x The column in data to map to the x-axis.
y The column in data to map to the y-axis.
size The column in data to map to the size of the points.
col The column in data to map to the color of the points.
... Additional aesthetic mappings or other arguments passed to ggplot.
Returns: A ggplot object representing the bubble plot.

Method `barchart_divergence()`: Create a divergence bar chart

This method generates a divergence bar chart where bars are colored based on their positive or negative value.

Usage:

```
Artist$barchart_divergence(data = self$data, group, y, fill, ...)
```

Arguments:

data A data frame containing the data to be plotted.
group The column in data representing the grouping variable.
y The column in data to map to the y-axis.
fill The column in data to map to the fill color of the bars.
... Additional aesthetic mappings or other arguments passed to ggplot.
Returns: A ggplot object representing the divergence bar chart.

Method `lollipop()`: Create a lollipop plot

This method generates a lollipop plot, where points are connected to a baseline by vertical segments, with customizable colors and labels.

Usage:

```
Artist$lollipop(data = self$data, x, y, ...)
```

Arguments:

data A data frame containing the data to be plotted.
x The column in data to map to the x-axis.
y The column in data to map to the y-axis.
... Additional aesthetic mappings or other arguments passed to ggplot.
Returns: A ggplot object representing the lollipop plot.

Method `contour()`: Create a contour plot

This method generates a contour plot that includes filled and outlined density contours, with data points overlaid.

Usage:

```
Artist$contour(data = self$data, x, y, ...)
```

Arguments:

data A data frame containing the data to be plotted.
x The column in data to map to the x-axis.
y The column in data to map to the y-axis.

... Additional aesthetic mappings or other arguments passed to ggplot.

Returns: A ggplot object representing the contour plot.

Method `scatter_ellipses()`: Create a scatter plot with ellipses

This method generates a scatter plot where data points are colored by group, with ellipses representing the confidence intervals for each group.

Usage:

```
Artist$scatter_ellipses(data = self$data, x, y, col, ...)
```

Arguments:

`data` A data frame containing the data to be plotted.

`x` The column in data to map to the x-axis.

`y` The column in data to map to the y-axis.

`col` The column in data to map to the color aesthetic.

... Additional aesthetic mappings or other arguments passed to ggplot.

Returns: A ggplot object representing the scatter plot with ellipses.

Method `donut()`: Create a donut plot

This method generates a donut plot, which is a variation of a pie chart with a hole in the center. The sections of the donut represent the proportion of categories in the data.

Usage:

```
Artist$donut(data = self$data, x, y, fill, ...)
```

Arguments:

`data` A data frame containing the data to be plotted.

`x` The column in data to map to the x-axis.

`y` The column in data to map to the y-axis.

`fill` The column in data to map to the fill color of the sections.

... Additional aesthetic mappings or other arguments passed to ggplot.

Returns: A ggplot object representing the donut plot.

Method `pie()`: Create a pie chart

This method generates a pie chart where sections represent the proportion of categories in the data.

Usage:

```
Artist$pie(data = self$data, y, fill, ...)
```

Arguments:

`data` A data frame containing the data to be plotted.

`y` The column in data to map to the y-axis.

`fill` The column in data to map to the fill color of the sections.

... Additional aesthetic mappings or other arguments passed to ggplot.

Returns: A ggplot object representing the pie chart.

Method `clone()`: The objects of this class are cloneable with this method.

Usage:

```
Artist$clone(deep = FALSE)
```

Arguments:

`deep` Whether to make a deep clone.

Examples

```
library(data.table)
air <- subset(airquality, Month %in% c(5, 6))
setDT(air)
cying <- Artist$new(data = air)
cying$plot_scatter(x = Wind, y = Temp)
cying$test_wilcox(
  formula = Ozone ~ Month,
)
cying$plot_scatter(x = Wind, y = Temp)
cying$plot_scatter(f = \"(x) x[, z := Wind * Temp], x = Wind, y = z)
```

available_ele

*Extract Unique Elements from a Column with Optional Filtering***Description**

Retrieves the unique, non-missing values from a specified column of a data frame. An optional expression can be provided to filter the rows of the data frame before extracting the values.

Usage

```
available_ele(data, col_name, subset)
```

Arguments

data	A data frame from which to extract values.
col_name	A single string specifying the name of the target column.
subset	An optional logical expression used to subset the data frame. This expression is evaluated in the context of the data, so columns can be referred to by their names directly (e.g., <code>Sepal.Length > 5</code>).

Value

A vector containing the unique, non-NA values from the specified column after the optional filtering has been applied.

Examples

```
# Example 1: Get all unique species from the iris dataset
available_ele(iris, "Species")

# Example 2: Get unique species for flowers with Sepal.Length > 7
available_ele(iris, "Species", subset = Sepal.Length > 7)

# Example 3: Get unique carb values for cars with 6 cylinders
available_ele(mtcars, "carb", subset = cyl == 6)
```

`available_tissue_class`*Retrieve Available Tissue Classes for a Given Species*

Description

This function extracts and returns a unique list of available tissue classes from the CellMarker2.0 database for a specified species.

Usage

```
available_tissue_class(spc)
```

Arguments

`spc` A character string specifying the species (e.g., "Human" or "Mouse").

Value

A character vector of unique tissue classes available for the given species. If no tissue classes are found, an empty vector is returned.

See Also

[available_tissue_type](#), [get_marker](#)

Examples

```
# Get all tissue classes for Human
available_tissue_class("Human")
```

`available_tissue_type` *Retrieve Available Tissue Types for a Given Species*

Description

This function extracts and returns a unique list of available tissue types from the CellMarker2.0 database for a specified species.

Usage

```
available_tissue_type(spc)
```

Arguments

`spc` A character string specifying the species (e.g., "Human" or "Mouse").

Value

A character vector of unique tissue types available for the given species. If no tissue types are found, an empty vector is returned.

See Also

[available_tissue_class](#), [get_marker](#)

Examples

```
# Get all tissue types for Human
available_tissue_type("Human")
```

check_marker	<i>Verify and Explore Cell Type Annotations</i>
--------------	---

Description

A post-analysis function that helps to verify and explore the automated cell type annotations generated by `matchCellMarker2`. It retrieves marker genes for the top-matching cell types of specified clusters, allowing for deeper inspection of the annotation results.

Usage

```
check_marker(marker, cl = c(), topcellN = 2, cis = FALSE)
```

Arguments

marker	A <code>data.table</code> object, which is the result of a call to <code>matchCellMarker2()</code> . This object must contain the attributes set by <code>matchCellMarker2</code> for the function to work correctly.
cl	A numeric or character vector specifying the cluster IDs to be inspected.
topcellN	An integer. For each cluster in <code>cl</code> , the function will retrieve markers for the top <code>topcellN</code> cell type annotations. Defaults to 2.
cis	A logical value that switches the function's mode. See Details. Defaults to <code>FALSE</code> .

Details

The function provides two distinct modes for marker retrieval, controlled by the `cis` parameter. This allows the user to answer two different, important questions:

- **cis = FALSE (Default): "Is the annotation correct?"** This mode answers the question by fetching the *canonical* markers for the annotated cell type from the reference database (via `get_marker`). It automatically uses the same filtering criteria (species, tissue, etc.) that were used in the original `matchCellMarker2` call, ensuring consistency.

- **cis = TRUE: "Why was this annotation made?"** This mode answers the question by extracting the *local* markers from the user's own data (i.e., the differentially expressed genes from the marker input) that led to the annotation. This helps understand the evidence behind the match.

Value

A named list. Each name in the list is a cell type, and each element is a character vector of its corresponding marker genes.

See Also

[matchCellMarker2](#) to generate the input for this function. [get_marker](#) which is used internally when `cis = FALSE`. [plotSeuratDot](#) to visualize the results.

Examples

```
## Not run:
library(easybio)
data(pbmcc.markers)

# Step 1: Generate cell type annotations
matched_cells <- matchCellMarker2(pbmcc.markers, n = 50, spc = "Human")

# Step 2: Verify the annotation for cluster 0.
# Let's check the top annotation (topcellN = 1).

# Question 1: "Is cluster 0 really a CD4-positive T cell?
# Let's see the canonical markers for it."
# Note: We don't need to pass 'spc' here; it's retrieved from matched_cells.
reference_markers <- check_marker(matched_cells, cl = 0, topcellN = 1)
print(reference_markers)
# Now you would typically use these markers in Seurat::DotPlot() or Seurat::FeaturePlot()

# Question 2: "Which of my genes made the algorithm think cluster 0
# is a CD4-positive T cell?"
local_markers <- check_marker(matched_cells, cl = 0, topcellN = 1, cis = TRUE)
print(local_markers)

## End(Not run)
```

CHOL_DEGs

Example DEGs data from Limma-Voom workflow for TCGA-CHOL project

Description

The data were obtained by the limma-voom workflow

dgeList	<i>Construct a DGEList Object</i>
---------	-----------------------------------

Description

This function creates a DGEList object from a count matrix, sample information, and feature information. It is designed to facilitate the analysis of differential gene expression using the edgeR package.

Usage

```
dgeList(count, sample.info, feature.info)
```

Arguments

count	A numeric matrix where rows represent features (e.g., genes) and columns represent samples. Row names should correspond to feature identifiers, and column names should correspond to sample identifiers.
sample.info	A data frame containing information about the samples. The number of rows should match the number of columns in the count matrix.
feature.info	A data frame containing information about the features. The number of rows should match the number of rows in the count matrix.

Value

A DGEList object as defined by the edgeR package, which includes the count data, sample information, and feature information.

dprocess_dgeList	<i>Filter Low-Expressed Genes and Normalize DGEList Data</i>
------------------	--

Description

This function filters out low-expressed genes from a DGEList object and normalizes the count data. It also provides diagnostic plots for raw and filtered data.

Usage

```
dprocess_dgeList(x, group.column, min.count = 10)
```

Arguments

<code>x</code>	A <code>DGEList</code> object containing raw count data.
<code>group.column</code>	The name of the column in <code>x\$samples</code> that contains the grouping information for the samples.
<code>min.count</code>	The minimum number of counts required for a gene to be considered expressed. Genes with counts below this threshold in any group will be filtered out. Defaults to 10.

Value

The function returns a `DGEList` object with low-expressed genes filtered out and normalization factors calculated.

finset

*Create a Vector from an Index-to-Label Map***Description**

Constructs a character vector by mapping labels to specified 0-based numeric indices. This is a utility function often used in single-cell analysis to assign cell type annotations to cluster IDs.

Usage

```
finset(
  x = list(c(0, 1, 3) ~ "Neutrophil", c(2, 4, 8) ~ "Macrophage"),
  len = integer(),
  setname = TRUE,
  na = "Unknown"
)
```

Arguments

<code>x</code>	The mapping of indices to labels. This can be provided in two formats: - A list of formulas, e.g., <code>list(c(0, 1) ~ "LabelA", 2 ~ "LabelB")</code>. - An expression object, e.g., <code>expression(c(0, 1) == "LabelA", 2 == "LabelB")</code>.
<code>len</code>	An optional integer specifying the minimum length of the output vector. If the highest index in <code>x</code> is greater than <code>len</code> , the vector will be automatically extended.
<code>setname</code>	A logical value. If <code>TRUE</code> (the default), the elements of the output vector are named with their corresponding 0-based index (e.g., "0", "1", "2", ...).
<code>na</code>	The character value used to fill positions that are not specified in the mapping. Defaults to "Unknown".

Value

A character vector with the specified labels at the given positions. The vector is named with 0-based indices if `setname` is `TRUE`.

Examples

```
# --- Example 1: Using the default formula list format ---
# This is the recommended and default usage.
mapping_formula <- list(
  c(0, 1, 3) ~ "Neutrophil",
  c(2, 4, 8) ~ "Macrophage"
)
fininsert(mapping_formula)

# --- Example 2: Using the expression format for backward compatibility ---
mapping_expr <- expression(
  c(0, 1, 3) == "Neutrophil",
  c(2, 4, 8) == "Macrophage"
)
fininsert(mapping_expr, len = 10, na = "Unassigned")
```

get_attr

Retrieve Attributes from an R Object

Description

This function extracts a specified attribute from an R object.

Usage

```
get_attr(x, attr_name)
```

Arguments

x	An R object that has attributes.
attr_name	The name of the attribute to retrieve.

Value

The value of the attribute with the given name.

get_marker

*Retrieve Markers for Specific Cells from cellMarker2***Description**

This function extracts a list of markers for one or more cell types from the cellMarker2 dataset. It allows filtering by species, cell type, the number of markers to retrieve, and a minimum count threshold for marker occurrences.

Usage

```
get_marker(
  spc,
  cell = character(),
  tissueClass = available_tissue_class(spc),
  tissueType = available_tissue_type(spc),
  number = 5,
  min.count = 1
)
```

Arguments

spc	A character string specifying the species, which can be either 'Human' or 'Mouse'.
cell	A character vector of cell types for which to retrieve markers.
tissueClass	A character specifying the tissue classes, default available_tissue_class(spc).
tissueType	A character specifying the tissue types, default available_tissue_type(spc).
number	An integer specifying the number of top markers to return for each cell type.
min.count	An integer representing the minimum number of times a marker must have been reported to be included in the results.

Value

A named list where each name corresponds to a cell type and each element is a vector of marker names.

Examples

```
# Example usage:
# Retrieve the top 5 markers for 'Macrophage' and 'Monocyte' cell types in humans,
# with a minimum count of 1.
library(easybio)
markers <- get_marker(spc = "Human", cell = c("Macrophage", "Monocyte"))
print(markers)
# Example with a typo in cell name
markers_typo <- get_marker(spc = "Human", cell = c("Macrophae", "Monocyte"))
```

groupStat

*Perform Summary Analysis by Group Using Regular Expressions***Description**

This function applies a specified function to each group defined by a regular expression pattern applied to the names of a data object. It is useful for summarizing data when groups are defined by a pattern in the names rather than a specific column or index.

Usage

```
groupStat(f, x, xname = colnames(x), patterns)
```

Arguments

f	A function that takes a single argument and returns a summary of the data.
x	A data frame or matrix containing the data to be summarized.
xname	A character vector containing the names of the variables in x.
patterns	A list of regular expressions that define the groups.

Value

A list containing the summary statistics for each group.

Examples

```
library(easybio)
groupStat(f = \(x) x + 1, x = mtcars, patterns = list("mp", "t"))
```

groupStatI

*Perform Summary Analysis by Group Using an column Index***Description**

This function applies a specified function to each group defined by an column index, and returns a summary of the results. It is useful for summarizing data by group when the groups are defined by an column index.

Usage

```
groupStatI(f, x, idx)
```


Arguments

f	A function that takes a single argument and returns a summary of the data.
x	A data frame or matrix containing the data to be summarized.
idx	A list of indices or group names that define the column groups.

Value

A list containing the summary statistics for each group.

Examples

```
library(easybio)
groupStatI(f = \(x) x + 1, x = mtcars, idx = list(c(1, 10), 2))
```

limmaFit*Fit a Linear Model for RNA-seq data using limma*

Description

This function fits a linear model to processed `DGEList` data using the `limma` package. It defines contrasts between groups and performs differential expression analysis.

Usage

```
limmaFit(x, group.column)
```

Arguments

x	A processed <code>DGEList</code> object containing normalized count data.
group.column	The name of the column in <code>x\$samples</code> that contains the grouping information for the samples.

Value

An eBayes object containing the fitted linear model and results of the differential expression analysis.

list2dt	<i>Convert a List with Vector Values to a Long Data.table</i>
---------	---

Description

This function converts a named list with vector values in each element to a long data.table. The list is first flattened into a single vector, and then the data.table is created with two columns: one for the name of the original list element and another for the value.

Usage

```
list2dt(x, col_names = c("name", "value"))
```

Arguments

x	A named list where each element contains a vector of values.
col_names	The colnames of the returned result.

Value

A long data.table with two columns: 'name' and 'value'.

Examples

```
library(easybio)
list2dt(list(a = c(1, 1), b = c(2, 2)))
```

list2graph	<i>Convert a Named List into a Graph Based on Overlap</i>
------------	---

Description

This function creates a graph from a named list, where the edges are determined by the overlap between the elements of the list. Each node in the graph represents an element of the list, and the weight of the edge between two nodes is the number of overlapping elements between the two corresponding lists.

Usage

```
list2graph(nodes)
```

Arguments

nodes	A named list where each element is a vector.
-------	--

Value

A data.table representing the graph, with columns for the node names (node_1 and node_2) and the weight of the edge (interWeight).

matchCellMarker2	<i>Annotate Clusters by Matching Markers with the CellMarker2.0 Database</i>
------------------	--

Description

This function takes cluster-specific markers, typically from `Seurat::FindAllMarkers`, and annotates each cluster with potential cell types by matching these markers against a reference database. It first filters and selects the top `n` marker genes for each cluster based on specified thresholds and then compares them to the reference database to find the most likely cell type annotations.

Usage

```
matchCellMarker2(
  marker,
  n,
  avg_log2FC_threshold = 0,
  p_val_adj_threshold = 0.05,
  spc,
  tissueClass = available_tissue_class(spc),
  tissueType = available_tissue_type(spc),
  ref = NULL
)
```

Arguments

marker	A data.frame or data.table of markers, usually the output of <code>Seurat::FindAllMarkers</code> . It must contain columns for cluster, gene, avg_log2FC, and p_val_adj.
n	An integer specifying the number of top marker genes to use from each cluster for matching. Genes are ranked by avg_log2FC after filtering.
avg_log2FC_threshold	A numeric value setting the minimum average log2 fold change for a marker to be considered. Defaults to 0.
p_val_adj_threshold	A numeric value setting the maximum adjusted p-value for a marker to be considered. Defaults to 0.05.
spc	A character string specifying the species, either "Human" or "Mouse". This is used to filter the cellMarker2 database. This parameter is ignored if a custom ref is provided.
tissueClass	A character vector of tissue classes to include from the cellMarker2 database. Defaults to all available tissue classes for the specified species. This parameter is ignored if a custom ref is provided. See <code>available_tissue_class()</code> .

tissueType	A character vector of tissue types to include from the cellMarker2 database. Defaults to all available tissue types for the specified species. This parameter is ignored if a custom ref is provided. See <code>available_tissue_type()</code> .
ref	An optional long data.frame which must contain 'cell_name' and 'marker' columns to be used as the reference for marker matching. If NULL (the default), the function uses the built-in cellMarker2 dataset. When a custom ref is provided, the spc, tissueClass, and tissueType parameters are ignored for the matching process itself, but their original values are saved for provenance.

Value

A data.table where each row represents a potential cell type match for a cluster. The table is keyed by cluster and includes columns for cluster, cell_name, uniqueN (number of unique matching markers), N (total matches), ordered_symbol (matching genes, ordered by frequency), and orderN (their frequencies).

The returned object also contains important attributes for downstream analysis:

ref	The reference data (either from cellMarker2 or the custom ref) used for the annotation.
is_custom_ref	A logical flag indicating if a custom ref was used.
filter_args	A list containing the filtering parameters used during the annotation, which is essential for the check_marker function.

See Also

[check_marker](#), [plotPossibleCell](#), [available_tissue_class](#), [available_tissue_type](#)

Examples

```
## Not run:
library(easybio)
data(pbmc.markers)

# Basic usage: Annotate clusters using the top 50 markers per cluster
matched_cells <- matchCellMarker2(pbmc.markers, n = 50, spc = "Human")
print(matched_cells)

# To see the top annotation for each cluster
top_matches <- matched_cells[, .SD[1], by = cluster]
print(top_matches)

# Advanced usage: Stricter filtering and focus on specific tissues
matched_cells_strict <- matchCellMarker2(
  pbmc.markers,
  n = 30,
  spc = "Human",
  avg_log2FC_threshold = 0.5,
  p_val_adj_threshold = 0.01,
  tissueType = c("Blood", "Bone marrow")
)
```

```

print(matched_cells_strict)

# --- Example with a custom reference ---
# Create a custom reference as a named list.
custom_ref_list <- list(
  "T-cell" = c("CD3D", "CD3E"),
  "B-cell" = c("CD79A", "MS4A1"),
  "Myeloid" = "LYZ"
)

# Convert the list to a long data.frame compatible with the 'ref' parameter.
custom_ref_df <- list2dt(custom_ref_list, col_names = c("cell_name", "marker"))

# Run annotation using the custom reference.
# When 'ref' is provided, the internal cellMarker2 database and its filters
# ('spc', 'tissueClass', 'tissueType') are ignored for matching.
matched_custom <- matchCellMarker2(
  pbmc.markers,
  n = 50,
  ref = custom_ref_df
)
print(matched_custom)

## End(Not run)

```

pbmc.markers

Example marker data from Seurat::FindAllMarkers()

Description

The data were obtained by the seurat PBMC workflow. exact script for this data is available as `system.file("example-single-cell.R", package="easybio")`

plotEnrichment2

Plot Enrichment for a Specific Pathway in fgsea

Description

This function creates a plot of enrichment scores for a specified pathway. It provides a visual representation of the enrichment score (ES) along with the ranks and ticks indicating the GSEA walk length.

Usage

```
plotEnrichment2(pathways, pwayname, stats, gseaParam = 1, ticksSize = 0.2)
```

Arguments

pathways	A list of pathways.
pwayname	The name of the pathway for which to plot enrichment.
stats	A rank vector obtained from the 'fgsea' package.
gseaParam	The GSEA walk length parameter. Default is 1.
ticksSize	The size of the tick marks. Default is 0.2.

Value

A ggplot object representing the enrichment plot.

plotGSEA	<i>Visualization of GSEA Result from <code>fgsea::fgsea()</code></i>
----------	--

Description

The plotGSEA function visualizes the results of a GSEA (Gene Set Enrichment Analysis) using data from the fgsea package. It generates a composite plot that includes an enrichment plot and a ranked metric plot.

Usage

```
plotGSEA(fgseaRes, pathways, pwayname, stats, save = FALSE)
```

Arguments

fgseaRes	A data table containing the GSEA results from the fgsea package.
pathways	A list of all pathways used in the GSEA analysis.
pwayname	The name of the pathway to visualize.
stats	A numeric vector representing the ranked statistics.
save	A logical value indicating whether to save the plot as a PDF file. Default is FALSE.

Value

ggplot2 object.

`plotMarkerDistribution`*Plot Distribution of a Marker Across Tissues and Cell Types*

Description

This function creates a dot plot displaying the distribution of a specified marker across different tissues and cell types, based on data from the CellMarker2.0 database.

Usage

```
plotMarkerDistribution(mkr = character())
```

Arguments

`mkr` character, the name of the marker to be plotted.

Value

A ggplot2 object representing the distribution of the marker.

Examples

```
## Not run:  
plotMarkerDistribution("CD14")  
  
## End(Not run)
```

`plotORA`*Visualization of ORA Test Results*

Description

The plotORA function visualizes the results of an ORA (Over-Representation Analysis) test. It generates a plot with customizable aesthetics for x, y, point size, and fill, with an option to flip the axes.

Usage

```
plotORA(data, x, y, size, fill, flip = FALSE)
```

Arguments

data	A data frame containing the ORA results to be visualized.
x	The column in data to map to the x-axis.
y	The column in data to map to the y-axis.
size	The column in data to map to the size of the points.
fill	The column in data to map to the fill color of the bars or points. Use a constant value for a single category.
flip	A logical value indicating whether to flip the axes of the plot. Default is FALSE.

Value

ggplot2 object.

plotPossibleCell	<i>Plot Possible Cell Distribution Based on matchCellMarker2() Results</i>
------------------	--

Description

This function creates a dot plot to visualize the distribution of possible cell types based on the results from the matchCellMarker2() function, utilizing data from the CellMarker2.0 database.

Usage

```
plotPossibleCell(marker, min.uniqueN = 2)
```

Arguments

marker	data.table, the result from the matchCellMarker2() function.
min.uniqueN	integer, the minimum number of unique marker genes that must be matched for a cell type to be included in the plot. Default is 2.

Value

A ggplot2 object representing the distribution of possible cell types.

plotRank

*Visualization of GSEA Rank Statistics***Description**

The plotRank function visualizes the ranked statistics of a GSEA (Gene Set Enrichment Analysis) analysis. The function creates a plot where the x-axis represents the rank of each gene, and the y-axis shows the corresponding ranked list metric.

Usage

```
plotRank(stats)
```

Arguments

stats A numeric vector containing the ranked statistics from a GSEA analysis.

Value

ggplot2 object

plotSeuratDot

*Create a Dot Plot to Visualize Marker Gene Expression***Description**

This function generates a `Seurat::DotPlot` to visualize the expression of specified marker genes across different cell clusters or groups. It is designed to work with a list of features, such as the output from the `check_marker` function.

Usage

```
plotSeuratDot(features, srt, split = FALSE, ...)
```

Arguments

features A named list of character vectors. Each name in the list represents a cell type or category, and the corresponding character vector contains the marker genes to be plotted for that category. This is typically the output of `check_marker()`.

srt A Seurat object containing the single-cell expression data.

split Logical, if TRUE, generates separate dot plots for each cell type in features

... Additional arguments passed to `Seurat::DotPlot()`, such as `cols`, `dot.scale`, etc.

Value

A ggplot2 object representing the dot plot, which can be further customized.

See Also

[check_marker](#) to generate the features list.

Examples

```
## Not run:
library(easybio)
library(Seurat)
data(pbmc.markers)

# In a real scenario, 'srt' would be your fully processed Seurat object.
# For this example, we create a minimal Seurat object.
# The expression matrix should contain the marker genes for the plot to be meaningful.
marker_genes <- unique(pbmc.markers$gene)
counts <- matrix(
  abs(rnorm(length(marker_genes) * 50, mean = 1, sd = 2)),
  nrow = length(marker_genes),
  ncol = 50
)
rownames(counts) <- marker_genes
colnames(counts) <- paste0("cell_", 1:50)

srt <- CreateSeuratObject(counts = counts)
srt$seurat_clusters <- sample(0:3, 50, replace = TRUE)
Idents(srt) <- "seurat_clusters"

# Step 1: Generate cell type annotations
matched_cells <- matchCellMarker2(pbmc.markers, n = 50, spc = "Human")

# Step 2: Get canonical markers for cluster 0's top annotation
reference_markers <- check_marker(matched_cells, cl = 0, topcellN = 1)

# Step 3: Plot the expression of these markers
if (!is.null(reference_markers) && length(reference_markers) > 0) {
  plotSeuratDot(features = reference_markers, srt = srt)
}

## End(Not run)
```

plotVolcano

Plot Volcano Plot for Differentially Expressed Genes

Description

This function generates a volcano plot for differentially expressed genes (DEGs) using ggplot2. It allows for customization of the plot with different aesthetic parameters.

Usage

```
plotVolcano(data, data.text, x, y, color, label)
```

Arguments

data	A data frame containing the DEGs result.
data.text	A data frame containing labeled data for text annotation.
x	variable representing the aesthetic for the x-axis.
y	variable representing the aesthetic for the y-axis.
color	variable representing the column name for the color aesthetic.
label	variable representing the column name for the text label aesthetic.

Value

A ggplot object representing the volcano plot.

prepare_geo

Download and Process GEO Data

Description

This function downloads gene expression data from the Gene Expression Omnibus (GEO) database. It retrieves either the expression matrix or the supplementary tabular data if the expression data is not available. The function also allows for the conversion of probe identifiers to gene symbols and can combine multiple probes into a single symbol.

Usage

```
prepare_geo(geo, dir = ".", combine = TRUE, method = "max")
```

Arguments

geo	A character string specifying the GEO Series ID (e.g., "GSE12345").
dir	A character string specifying the directory where files should be downloaded. Default is the current working directory (".").
combine	A logical value indicating whether to combine multiple probes into a single gene symbol. Default is TRUE.
method	A character string specifying the method to use for combining probes into a single gene symbol. Options are "max" (take the maximum value) or "mean" (compute the average). Default is "max".

Value

A list containing:

data	A data frame of the expression matrix.
sample	A data frame of the sample metadata.
feature	A data frame of the feature metadata, which includes gene symbols if combining probes.

prepare_tcga	<i>Prepare TCGA Data for Analysis</i>
--------------	---------------------------------------

Description

This function prepares TCGA data for downstream analyses such as differential expression analysis with limma or survival analysis. It extracts and processes the necessary information from the TCGA data object, separating tumor and non-tumor samples.

Usage

```
prepare_tcga(data)
```

Arguments

data	A SummarizedExperiment object containing TCGA data, typically obtained from R package TCGABiolinks.
------	---

Value

A list.

setcolnames	<i>Rename Column Names of a Data Frame or Matrix</i>
-------------	--

Description

This function renames the column names of a data frame or matrix to the specified names.

Usage

```
setcolnames(object, nm)
```

Arguments

object	A data frame or matrix whose column names will be renamed.
nm	A character vector containing the new names for the columns.

Value

A data frame or matrix with the new column names.

setrownames

Rename Row Names of a Data Frame or Matrix

Description

This function renames the row names of a data frame or matrix to the specified names.

Usage

```
setrownames(object, nm)
```

Arguments

object	A data frame or matrix whose row names will be renamed.
nm	A character vector containing the new names for the rows.

Value

A data frame or matrix with the new row names.

setSavendir

Set a Directory for Saving Files

Description

This function sets a directory path for saving files, creating the directory if it does not already exist. The directory path is created with the given arguments, which are passed directly to `file.path()`.

Usage

```
setSavendir(...)
```

Arguments

...	Arguments to be passed to <code>file.path()</code> to construct the directory path.
-----	---

Value

The path to the newly created or existing directory.

split_matrix	<i>Split a Matrix into Smaller Sub-matrices by Column or Row</i>
--------------	--

Description

This function splits a matrix into multiple smaller matrices by column or row. It is useful for processing large matrices in chunks, such as when performing analysis on a single computer with limited memory.

Usage

```
split_matrix(matrix, chunk_size, column = TRUE)
```

Arguments

matrix	A numeric or logical matrix to be split.
chunk_size	The number of columns or rows to include in each smaller matrix.
column	Divided by column(default is TRUE)

Value

A list of smaller matrices, each with chunk_size columns or rows.

Examples

```
library(easybio)
split_matrix(mtcars, chunk_size = 2)
split_matrix(mtcars, chunk_size = 5, column = FALSE)
```

suggest_best_match	<i>Suggest Best Matches for a String from a Vector of Choices</i>
--------------------	---

Description

This function provides intelligent suggestions for a user's input string by finding the best matches from a given vector of choices. It follows a multi-layered approach:

1. Performs normalization (case-insensitivity, trimming whitespace).
2. Checks for an exact match first for maximum performance and accuracy.
3. If no exact match, it uses a combination of fuzzy string matching (Levenshtein distance via `adist`) to catch typos and partial/substring matching (`grep`) to handle incomplete input.
4. Ranks the potential matches and returns the best suggestion(s).

Usage

```
suggest_best_match(
  x,
  choices,
  n = 1,
  threshold = 2,
  ignore.case = TRUE,
  return_distance = FALSE
)
```

Arguments

x	A single character string; the user input to find matches for.
choices	A character vector of available, valid options.
n	An integer specifying the maximum number of suggestions to return. Defaults to 1.
threshold	An integer; the maximum Levenshtein distance to consider a choice a "close" match. A lower value is stricter. Defaults to 2.
ignore.case	A logical value. If TRUE, matching is case-insensitive. Defaults to TRUE.
return_distance	A logical value. If TRUE, the output is a data.frame containing the suggestions and their calculated distance/score. Defaults to FALSE.

Value

By default (`return_distance = FALSE`), returns a character vector of the best `n` suggestions. If no suitable match is found, returns NA. If `return_distance = TRUE`, returns a data.frame with columns `suggestion` and `distance`, or NULL if no match is found.

Examples

```
# --- Setup ---
cell_types <- c(
  "B cell", "T cell", "Macrophage", "Monocyte", "Neutrophil",
  "Natural Killer T-cell", "Dendritic cell"
)

# --- Usage ---
# 1. Exact match (after normalization)
suggest_best_match("t cell", cell_types)
#> [1] "T cell"

# 2. Typo correction (fuzzy match)
suggest_best_match("Macrophag", cell_types)
#> [1] "Macrophage"

# 3. Partial input (substring match)
suggest_best_match("Mono", cell_types)
#> [1] "Monocyte"
```

```
# 4. Requesting multiple suggestions
suggest_best_match("t", cell_types, n = 3)
#> [1] "T cell" "Neutrophil" "Natural Killer T-cell"

# 5. No good match found
suggest_best_match("Erythrocyte", cell_types)
#> [1] NA

# 6. Returning suggestions with their distance score
suggest_best_match("t ce", cell_types, n = 3, return_distance = TRUE)
#>          suggestion distance
#> 1          T cell          1
#> 2      Dendritic cell          2
#> 3 Natural Killer T-cell          2
```

theme_publication	<i>Custom ggplot2 Theme for Academic Publications</i>
-------------------	---

Description

theme_publication creates a custom ggplot2 theme designed for academic publications, ensuring clarity, readability, and a professional appearance. It is based on theme_classic() and includes additional refinements to axis lines, text, and other plot elements to meet the standards of high-quality academic figures.

Usage

```
theme_publication(base_size = 12, base_family = "sans")
```

Arguments

base_size	numeric, the base font size. Default is 12.
base_family	character, the base font family. Default is "sans".

Value

A ggplot2 theme object that can be applied to ggplot2 plots.
ggplot2 theme.

Examples

```
library(ggplot2)
p <- ggplot(mtcars, aes(mpg, wt)) +
  geom_point() +
  theme_publication()
print(p)
```

tuneParameters	<i>Optimize Resolution and Gene Number Parameters for Cell Type Annotation</i>
----------------	--

Description

This function tunes the resolution parameter in `Seurat::FindClusters()` and the number of top differential genes (N) to obtain different cell type annotation results. The function generates UMAP plots for each parameter combination, allowing for a comparison of how different settings affect the clustering and annotation.

Usage

```
tuneParameters(srt, resolution = numeric(), N = integer(), spc)
```

Arguments

srt	Seurat object, the input data object to be analyzed.
resolution	numeric vector, a vector of resolution values to be tested in <code>Seurat::FindClusters()</code> .
N	integer vector, a vector of values indicating the number of top differential genes to be used for matching in <code>matchCellMarker2()</code> .
spc	character, the species parameter for the <code>matchCellMarker2()</code> function, specifying the organism.

Value

A list of ggplot2 objects, each representing a UMAP plot generated with a different combination of resolution and N parameters.

uniprot_id_map	<i>Map UniProt IDs to Other Identifiers</i>
----------------	---

Description

This function maps UniProt IDs to other identifiers using UniProt's ID mapping service. It sends a request to the UniProt API to perform the mapping and retrieves the results in a tabular format.

Usage

```
uniprot_id_map(...)
```

Arguments

...	Parameters to be passed in the request body.
-----	--

Value

A data.table containing the mapped identifiers.

Examples

```
## Not run:
uniprot_id_map(
  ids = "P21802,P12345",
  from = "UniProtKB_AC-ID",
  to = "UniRef90"
)

## End(Not run)
```

workIn	<i>Perform Operations in a Specified Directory and Return to the Original Directory</i>
--------	---

Description

This function allows you to perform operations in a specified directory and then return to the original directory. It is useful when you need to work with files or directories that are located in a specific location, but you want to return to the original working directory after the operation is complete.

Usage

```
workIn(dir, expr)
```

Arguments

dir	The directory path in which to operate. If the directory does not exist, it will be created recursively.
expr	An R expression to be evaluated within the specified directory.

Value

The result of evaluating the expression within the specified directory.

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