Package ‘tidybulk’

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Type Package
Title Brings transcriptomics to the tidyverse
Version 1.14.3
Description This is a collection of utility functions that allow to perform exploration of and calculations to RNA sequencing data, in a modular, pipe-friendly and tidy fashion.
License GPL-3
Depends R (>= 4.1.0), ttservice (>= 0.3.6)
Imports tibble, readr, dplyr (>= 1.1.0), magrittr, tidyr, stringr, stringi, purrr, tidyselect, preprocessCore, stats, parallel, utils, lifecycle, scales, SummarizedExperiment, GenomicRanges, methods, S4Vectors, crayon, Matrix
Suggests BiocStyle, testthat, vctrs, AnnotationDbi, BiocManager, Rsubearead, e1071, edgeR, limma, org.Hs.eg.db, org姚Mm.eg.db, sva, GGally, knitr, qpdf, covr, Seurat, KernSmooth, Rtsne, ggplot2, widyr, clusterProfiler, msigdbR, DESeq2, broom, survival, boot, betareg, tidyHeatmap, pasilla, ggrepel, devtools, functional, survminer, tidySummarizedExperiment, markdown, uwot, matrixStats, igraph, EGSEA, IRanges, here, glmnSeq, pbapply, pbmcapply, lme4, glmmTMB, MASS, pkgconfig
VignetteBuilder knitr
RdMacros lifecycle
Biarch true
biocViews AssayDomain, Infrastructure, RNASeq, DifferentialExpression, GeneExpression, Normalization, Clustering, QualityControl, Sequencing, Transcription, Transcriptomics
Encoding UTF-8
LazyData true
RoxygenNote 7.2.3
LazyDataCompression xz
URL https://github.com/stemangiola/tidybulk
**R topics documented:**

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adjust_abundance

Adjust transcript abundance for unwanted variation

Description

adjust_abundance() takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or a ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with an additional adjusted abundance column. This method uses scaled counts if present.
Usage

adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
  ...,
  log_transform = NULL,
  transform = NULL,
  inverse_transform = NULL
)

## S4 method for signature 'spec_tbl_df'
adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
  ...,
  log_transform = NULL,
  transform = NULL,
  inverse_transform = NULL
)

## S4 method for signature 'tbl_df'
adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
  ...,
  log_transform = NULL,
  transform = NULL,
  inverse_transform = NULL
)
### `adjust_abundance`

```r
inverse_transform = NULL
)

## S4 method for signature 'tidybulk'
adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
  ...
)

## S4 method for signature 'SummarizedExperiment'
adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
  ...
)

## S4 method for signature 'RangedSummarizedExperiment'
adjust_abundance(
  .data,
  .formula = NULL,
  .factor_unwanted = NULL,
  .factor_of_interest = NULL,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "combat_seq",
  action = "add",
)```
...,
  log_transform = NULL,
  transform = NULL,
  inverse_transform = NULL
)

Arguments
.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
.formula DEPRECATED - A formula with no response variable, representing the desired linear model where the first covariate is the factor of interest and the second covariate is the unwanted variation (of the kind ~ factor_of_interest + batch)
.factor_unwanted A tidy select, e.g. column names without double quotation. c(batch, country) These are the factor that we want to adjust for, including unwanted batch effect, and unwanted biological effects.
.factor_of_interest A tidy select, e.g. column names without double quotation. c(treatment) These are the factor that we want to preserve.
.sample The name of the sample column
.transcript The name of the transcript/gene column
.abundance The name of the transcript/gene abundance column
.method A character string. Methods include combat_seq (default), combat and limma_remove_batch_effect.
.action A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).
... Further parameters passed to the function sva::ComBat
.log_transform DEPRECATED - A boolean, whether the value should be log-transformed (e.g., TRUE for RNA sequencing data)
.transform DEPRECATED - A function that will tranform the counts, by default it is log1p for RNA sequencing data, but for avoiding tranformation you can use identity
.inverse_transform DEPRECATED - A function that is the inverse of transform (e.g. expm1 is inverse of log1p). This is needed to tranform back the counts after analysis.

Details

‘r lifecycle::badge("maturing")’

This function adjusts the abundance for (known) unwanted variation. At the moment just an unwanted covariate is allowed at a time using Combat (DOI: 10.1093/bioinformatics/bts034)

Underlying method: sva::ComBat(data, batch = my_batch, mod = design, prior.plots = FALSE, ...)

adjust_abundance
aggregate_duplicates

Value

A consistent object (to the input) with additional columns for the adjusted counts as '<COUNT COLUMN>_adjusted'

A consistent object (to the input) with additional columns for the adjusted counts as '<COUNT COLUMN>_adjusted'

A consistent object (to the input) with additional columns for the adjusted counts as '<COUNT COLUMN>_adjusted'

A consistent object (to the input) with additional columns for the adjusted counts as '<COUNT COLUMN>_adjusted'

A ‘SummarizedExperiment’ object

A ‘SummarizedExperiment’ object

Examples

```r
cm = tidybulk::se_mini
cm$batch = 0
cm$batch[colnames(cm) %in% c("SRR1740035", "SRR1740043") ] = 1
cm |> identify_abundant() |> adjust_abundance(.factor_unwanted = batch, .factor_of_interest = condition, method="combat"
```

aggregate_duplicates   Aggregates multiple counts from the same samples (e.g., from isoforms), concatenates other character columns, and averages other numeric columns

Description

aggregate_duplicates() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with aggregated transcripts that were duplicated.

Usage

```r
aggregate_duplicates(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
```

aggregate_duplicates

aggregate_duplicates(.data, .sample = NULL, .transcript = NULL, .abundance = NULL, aggregation_function = sum, keep_integer = TRUE)

## S4 method for signature 'spec_tbl_df'
agegregate_duplicates(.data, .sample = NULL, .transcript = NULL, .abundance = NULL, aggregation_function = sum, keep_integer = TRUE)

## S4 method for signature 'tbl_df'
agegregate_duplicates(.data, .sample = NULL, .transcript = NULL, .abundance = NULL, aggregation_function = sum, keep_integer = TRUE)

## S4 method for signature 'tidybulk'
agegregate_duplicates(.data, .sample = NULL, .transcript = NULL, .abundance = NULL, aggregation_function = sum, keep_integer = TRUE)

## S4 method for signature 'SummarizedExperiment'
agegregate_duplicates(.data, .sample = NULL, .transcript = NULL, .abundance = NULL, aggregation_function = sum, keep_integer = TRUE)

## S4 method for signature 'RangedSummarizedExperiment'
agegregate_duplicates(.data, .sample = NULL,
aggregate_duplicates

```r
.aggregate_duplicates = function(
  .transcript = NULL,
  .abundance = NULL,
  aggregation_function = sum,
  keep_integer = TRUE
)
```

**Arguments**

- `.data` A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
- `.sample` The name of the sample column
- `.transcript` The name of the transcript/gene column
- `.abundance` The name of the transcript/gene abundance column
- `aggregation_function` A function for counts aggregation (e.g., sum, median, or mean)
- `keep_integer` A boolean. Whether to force the aggregated counts to integer

**Details**

`r lifecycle::badge("maturing")`

This function aggregates duplicated transcripts (e.g., isoforms, ensembl). For example, we often have to convert ensembl symbols to gene/transcript symbol, but in doing so we have to deal with duplicates. `aggregate_duplicates` takes a tibble and column names (as symbols; for `sample`, `transcript` and `count`) as arguments and returns a tibble with aggregate transcript with the same name. All the rest of the column are appended, and factors and boolean are appended as characters.

Underlying custom method: `data |> filter(n_aggr > 1) |> group_by(!!.sample,!!.transcript) |> dplyr::mutate(!!.abundance := !!.abundance |> aggregation_function())`

**Value**

A consistent object (to the input) with aggregated transcript abundance and annotation

A consistent object (to the input) with aggregated transcript abundance and annotation

A consistent object (to the input) with aggregated transcript abundance and annotation

A consistent object (to the input) with aggregated transcript abundance and annotation

A `SummarizedExperiment` object

A `SummarizedExperiment` object

**Examples**

```r
# Create a aggregation column
se_mini = tidybulk::se_mini
SummarizedExperiment::rowData(se_mini )$gene_name = rownames(se_mini )
aggregate_duplicates(
  se_mini,
  .transcript = "transcript",
  .abundance = "abundance",
  aggregation_function = mean,
  keep_integer = TRUE
)
```

Description

‘arrange()’ order the rows of a data frame rows by the values of selected columns.

Unlike other dplyr verbs, ‘arrange()’ largely ignores grouping; you need to explicit mention grouping variables (or use ‘by_group = TRUE’) in order to group by them, and functions of variables are evaluated once per data frame, not once per group.

Arguments

- `.data` A data frame, data frame extension (e.g. a tibble), or a lazy data frame (e.g. from dbplyr or dtplyr). See *Methods*, below, for more details.
- `...` `<[\{tidy-eval\}[dpolyr_tidy_eval]> Variables, or functions or variables. Use [desc()] to sort a variable in descending order.
- `.by_group` If TRUE, will sort first by grouping variable. Applies to grouped data frames only.

Details

## Locales
The sort order for character vectors will depend on the collating sequence of the locale in use: see [locales()].

## Missing values
Unlike base sorting with ‘sort()’, ‘NA’ are: * always sorted to the end for local data, even when wrapped with ‘desc()’. * treated differently for remote data, depending on the backend.

Value

An object of the same type as ‘.data’.

* All rows appear in the output, but (usually) in a different place. * Columns are not modified. * Groups are not modified. * Data frame attributes are preserved.

A tibble

Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages:
See Also

Other single table verbs: `filter()`, `mutate()`, `rename()`, `summarise()`

Examples

```r
arrange(mtcars, cyl, disp)
```

---

### as_matrix

**Get matrix from tibble**

**Description**

Get matrix from tibble

**Usage**

```r
as_matrix(tbl, rownames = NULL, do_check = TRUE)
```

**Arguments**

- `tbl`: A tibble
- `rownames`: The column name of the input tibble that will become the rownames of the output matrix
- `do_check`: A boolean

**Value**

A matrix

**Examples**

```r
tibble(.feature = "CD3G", count=1) |> as_matrix(rownames=.feature)
```
as_SummarizedExperiment

Description

as_SummarizedExperiment() creates a ‘SummarizedExperiment’ object from a ‘tbl’ or ‘tidybulk’
tbl formatted as | <SAMPLE> | <TRANSCRIPT> | <COUNT> | <...> |

Usage

as_SummarizedExperiment(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL
)

## S4 method for signature 'spec_tbl_df'
as_SummarizedExperiment(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL
)

## S4 method for signature 'tbl_df'
as_SummarizedExperiment(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL
)

## S4 method for signature 'tidybulk'
as_SummarizedExperiment(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL
)

Arguments

.data A tibble
.sample The name of the sample column
bind_cols

Arguments

x        tbls to join. (See dplyr)
y        tbls to join. (See dplyr)
by        A character vector of variables to join by. (See dplyr)
copy        If x and y are not from the same data source, and copy is TRUE, then y will be copied into the same src as x. (See dplyr)
suffix        If there are non-joined duplicate variables in x and y, these suffixes will be added to the output to disambiguate them. Should be a character vector of length 2. (See dplyr)
...        Data frames to combine (See dplyr)

Value

A tt object

Examples

annotation = tidybulk::se_mini |> tidybulk() |> as_tibble() |> distinct(.sample) |> mutate(source = "AU")
tidybulk::se_mini |> tidybulk() |> as_tibble() |> left_join(annotation)
bind_rows

Efficiently bind multiple data frames by row and column

Description

This is an efficient implementation of the common pattern of `do.call(rbind, dfs)` or `do.call(cbind, dfs)` for binding many data frames into one.

Arguments

... Data frames to combine.

Each argument can either be a data frame, a list that could be a data frame, or a list of data frames.

When row-binding, columns are matched by name, and any missing columns will be filled with NA.

When column-binding, rows are matched by position, so all data frames must have the same number of rows. To match by value, not position, see mutatejoins.

.id Data frame identifier.

When `.id` is supplied, a new column of identifiers is created to link each row to its original data frame. The labels are taken from the named arguments to `bind_rows()`. When a list of data frames is supplied, the labels are taken from the names of the list. If no names are found a numeric sequence is used instead.

.add.cell.ids from Seurat 3.0 A character vector of length(x = c(x, y)). Appends the corresponding values to the start of each objects’ cell names.

Details

The output of `bind_rows()` will contain a column if that column appears in any of the inputs.

Value

`bind_rows()` and `bind_cols()` return the same type as the first input, either a data frame, `tbl_df`, or `grouped_df`.

Examples

data(se_mini)

se_mini_tidybulk = se_mini |> tidybulk()
bind_rows( se_mini_tidybulk, se_mini_tidybulk )

tt_bind = se_mini_tidybulk |> select(time, condition)
se_mini_tidybulk |> bind_cols(tt_bind)
Description

Needed for vignette breast_tcga_mini_SE

Usage

breast_tcga_mini_SE

Format

An object of class SummarizedExperiment with 500 rows and 251 columns.

check_if_counts_is_na

Description

Check whether there are NA counts

Usage

check_if_counts_is_na(.data, .abundance)

Arguments

.data A tibble of read counts
.abundance A character name of the read count column

Value

A tbl
check_if_duplicated_genes

Check whether there are duplicated genes/transcripts

Description

Check whether there are duplicated genes/transcripts

Usage

check_if_duplicated_genes(
  .data,
  .sample = sample,
  .transcript = transcript,
  .abundance = `read count`
)

Arguments

  .data A tibble of read counts
  .sample A character name of the sample column
  .transcript A character name of the transcript/gene column
  .abundance A character name of the read count column

Value

A tbl

check_if_wrong_input

Check whether there are NA counts

Description

Check whether there are NA counts

Usage

check_if_wrong_input(.data, list_input, expected_type)

Arguments

  .data A tibble of read counts
  list_input A list
  expected_type A character string
Value

A tbl

Description

cluster_elements() takes as input a 'tbl' (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and identify clusters in the data.

Usage

cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  transform = log1p,
  action = "add",
  ...,
  log_transform = NULL
)

## S4 method for signature 'spec_tbl_df'
cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  transform = log1p,
  action = "add",
  ...,
  log_transform = NULL
)

## S4 method for signature 'tbl_df'
cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
cluster_elements

.abundance = NULL,
method,
of_samples = TRUE,
transform = log1p,
action = "add",
...
log_transform = NULL
}

## S4 method for signature 'tidybulk'
cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
of_samples = TRUE,
  transform = log1p,
  action = "add",
  ...
log_transform = NULL
)

## S4 method for signature 'SummarizedExperiment'
cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
of_samples = TRUE,
  transform = log1p,
  action = "add",
  ...
log_transform = NULL
)

## S4 method for signature 'RangedSummarizedExperiment'
cluster_elements(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
of_samples = TRUE,
  transform = log1p,
  action = "add",
  ...
,
log_transform = NULL
)

Arguments

.data A 'tbl' (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
.element The name of the element column (normally samples).
.feature The name of the feature column (normally transcripts/genes)
.abundance The name of the column including the numerical value the clustering is based on (normally transcript abundance)
.method A character string. The cluster algorithm to use, at the moment k-means is the only algorithm included.
of_samples A boolean. In case the input is a tidybulk object, it indicates Whether the element column will be sample or transcript column
.transform A function that will transform the counts, by default it is log1p for RNA sequencing data, but for avoiding transformation you can use identity
.action A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).
... Further parameters passed to the function kmeans
.log_transform DEPRECATED - A boolean, whether the value should be log-transformed (e.g., TRUE for RNA sequencing data)

Details

‘r lifecycle::badge("maturing")’
identifies clusters in the data, normally of samples. This function returns a tibble with additional columns for the cluster annotation. At the moment only k-means (DOI: 10.2307/2346830) and SNN clustering (DOI:10.1016/j.cell.2019.05.031) is supported, the plan is to introduce more clustering methods.
Underlying method for kmeans do.call(kmeans(.data, iter.max = 1000, ...)
Underlying method for SNN .data Seurat::CreateSeuratObject() Seurat::ScaleData(display.progress = TRUE,num.cores = 4, do.par = TRUE) Seurat::FindVariableFeatures(selection.method = "vst") Seurat::RunPCA(npcs = 30) Seurat::FindNeighbors() Seurat::FindClusters(method = "igraph", ...)

Value

A tbl object with additional columns with cluster labels
A tbl object with additional columns with cluster labels
A tbl object with additional columns with cluster labels
A tbl object with additional columns with cluster labels
A ‘SummarizedExperiment’ object
A ‘SummarizedExperiment’ object
Examples

cluster_elements(tidybulk::se_mini, centers = 2, method="kmeans")

counts_ensembl

Counts with ensembl annotation

Description

Counts with ensembl annotation

Usage

counts_ensembl

Format

An object of class tbl_df (inherits from tbl, data.frame) with 119 rows and 6 columns.

deconvolve_cellularity

Get cell type proportions from samples

Description

deconvolve_cellularity() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with the estimated cell type abundance for each sample

Usage

deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
  ...
)
## S4 method for signature 'spec_tbl_df'
deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
  ...
)

## S4 method for signature 'tbl_df'
deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
  ...
)

## S4 method for signature 'tidybulk'
deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
  ...
)

## S4 method for signature 'SummarizedExperiment'
deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
deconvolve_cellularity

```r
action = "add",
...
)

## S4 method for signature 'RangedSummarizedExperiment'
deconvolve_cellularity(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  reference = NULL,
  method = "cibersort",
  prefix = "",
  action = "add",
  ...
)
```

### Arguments

- **.data**: A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
- **.sample**: The name of the sample column
- **.transcript**: The name of the transcript/gene column
- **.abundance**: The name of the transcript/gene abundance column
- **reference**: A data frame. The methods cibersort and llsr can accept a custom rectangular dataframe with genes as rows names, cell types as column names and gene-transcript abundance as values. For example tidybulk::X_cibersort. The transcript/cell_type data frame of integer transcript abundance. If NULL, the default reference for each algorithm will be used. For llsr will be LM22.
- **method**: A character string. The method to be used. At the moment Cibersort (default, can accept custom reference), epic (can accept custom reference) and llsr (linear least squares regression, can accept custom reference), mcp_counter, quantiseq, xcell are available.
- **prefix**: A character string. The prefix you would like to add to the result columns. It is useful if you want to reshape data.
- **action**: A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).
- **...**: Further parameters passed to the function Cibersort

### Details

'\texttt{r lifecycle::badge("maturing")}'

This function infers the cell type composition of our samples (with the algorithm Cibersort; Newman et al., 10.1038/nmeth.3337).

Underlying method: CIBERSORT(Y = data, X = reference, ...)
Value
A consistent object (to the input) including additional columns for each cell type estimated
A consistent object (to the input) including additional columns for each cell type estimated
A consistent object (to the input) including additional columns for each cell type estimated
A consistent object (to the input) including additional columns for each cell type estimated
A `SummarizedExperiment` object
A `SummarizedExperiment` object

Examples

```r
# Subsetting for time efficiency
tidybulk::se_mini |> deconvolve_cellularity(cores = 1)
```

Description
Get DESCRIPTION from gene SYMBOL for Human and Mouse

describe_transcript
describe_transcript
describe_transcript
describe_transcript
describe_transcript
describe_transcript
describe_transcript
describe_transcript

describe_transcript(.data, .transcript = NULL)

Usage

describe_transcript(.data, .transcript = NULL)
```
## S4 method for signature 'spec_tbl_df'
describe_transcript(.data, .transcript = NULL)
```
## S4 method for signature 'tbl_df'
describe_transcript(.data, .transcript = NULL)
```
## S4 method for signature 'tidybulk'
describe_transcript(.data, .transcript = NULL)
```
`.describe_transcript_SE(.data, .transcript = NULL)`
## S4 method for signature 'SummarizedExperiment'
describe_transcript(.data, .transcript = NULL)

## S4 method for signature 'RangedSummarizedExperiment'
describe_transcript(.data, .transcript = NULL)

### Arguments

- **.data**
  - A tt or tbl object.

- **.transcript**
  - A character. The name of the gene symbol column.

### Value

A tbl

A consistent object (to the input) including additional columns for transcript symbol

A consistent object (to the input) including additional columns for transcript symbol

A consistent object (to the input) including additional columns for transcript symbol

A ‘SummarizedExperiment’ object

A consistent object (to the input) including additional columns for transcript symbol

A consistent object (to the input) including additional columns for transcript symbol

### Examples

```r
describe_transcript(tidybulk::se_mini)
```

---

### distinct

#### Description

distinct

#### Arguments

- **.data**
  - A tbl. (See dplyr)

- **...**
  - Data frames to combine (See dplyr)

- **.keep_all**
  - If TRUE, keep all variables in .data. If a combination of ... is not distinct, this keeps the first row of values. (See dplyr)

#### Value

A tt object
Examples

```r
tidybulk::se_mini |> tidybulk() |> distinct()
```

---

### ensembl_symbol_mapping

**Data set**

**Description**

Data set

**Usage**

```r
ensembl_symbol_mapping
```

**Format**

An object of class `spec_tbl_df` (inherits from `tbl_df`, `tbl`, `data.frame`) with 291249 rows and 3 columns.

---

### ensembl_to_symbol

**Add transcript symbol column from ensembl id for human and mouse data**

**Description**

`ensembl_to_symbol()` takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with the additional transcript symbol column

**Usage**

```r
ensembl_to_symbol(.data, .ensembl, action = "add")
```

```r
## S4 method for signature 'spec_tbl_df'
ensembl_to_symbol(.data, .ensembl, action = "add")
```

```r
## S4 method for signature 'tbl_df'
ensembl_to_symbol(.data, .ensembl, action = "add")
```

```r
## S4 method for signature 'tidybulk'
ensembl_to_symbol(.data, .ensembl, action = "add")
```
fill_missing_abundance

Arguments

.data a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.ensembl A character string. The column that is represents ensembl gene id

action A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).

Details

[Questioning]
This is useful since different resources use ensembl IDs while others use gene symbol IDs. At the moment this work for human (genes and transcripts) and mouse (genes) data.

Value

A consistent object (to the input) including additional columns for transcript symbol
A consistent object (to the input) including additional columns for transcript symbol
A consistent object (to the input) including additional columns for transcript symbol
A consistent object (to the input) including additional columns for transcript symbol

Examples

# This function was designed for data.frame
# Convert from SummarizedExperiment for this example. It is NOT recoemended.
tidybulk::se_mini |> tidybulk() |> as_tibble() |> ensembl_to_symbol(.feature)

fill_missing_abundance

Fill transcript abundance if missing from sample-transcript pairs

Description

fill_missing_abundance() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with new observations
fill_missing_abundance

Usage

fill_missing_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  fill_with
)

## S4 method for signature 'spec_tbl_df'
fill_missing_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  fill_with
)

## S4 method for signature 'tbl_df'
fill_missing_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  fill_with
)

## S4 method for signature 'tidybulk'
fill_missing_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  fill_with
)

Arguments

.data A 'tbl' formatted as | <SAMPLE> | <TRANSCRIPT> | <COUNT> | <...>
.sample The name of the sample column
.transcript The name of the transcript column
.abundance The name of the transcript abundance column
.fill_with A numerical abundance with which fill the missing data points

Details

[Questioning]
This function fills the abundance of missing sample-transcript pair using the median of the sample group defined by the formula

**Value**

A consistent object (to the input) non-sparse abundance
A consistent object (to the input) with filled abundance
A consistent object (to the input) with filled abundance
A consistent object (to the input) with filled abundance

**Examples**

```r
print("Not run for build time.")

# tidybulk::se_mini |> fill_missing_abundance( fill_with = 0)
```

---

**Description**

`filter()` retains the rows where the conditions you provide a ‘TRUE’. Note that, unlike base sub-setting with ‘[’, rows where the condition evaluates to ‘NA’ are dropped.

**Arguments**

- `.data` A tbl. (See dplyr)
- `...` Logical predicates defined in terms of the variables in `.data`. Multiple conditions are combined with ‘&’. Only rows where the condition evaluates to ‘TRUE’ are kept.
- `.preserve` when ‘FALSE’ (the default), the grouping structure is recalculated based on the resulting data, otherwise it is kept as is.

**Details**

dplyr is not yet smart enough to optimise filtering optimisation on grouped datasets that don’t need grouped calculations. For this reason, filtering is often considerably faster on [ungroup()ed data.

**Value**

An object of the same type as `.data`.

* Rows are a subset of the input, but appear in the same order. * Columns are not modified. * The number of groups may be reduced (if `.preserve` is not ‘TRUE’). * Data frame attributes are preserved.
Useful filter functions

* ['=='], ['>'], ['>='] etc * ['&'], ['|'], ['^'], [xor()] * [is.na()] * [between()], [near()]

Grouped tibbles

Because filtering expressions are computed within groups, they may yield different results on grouped tibbles. This will be the case as soon as an aggregating, lagging, or ranking function is involved. Compare this ungrouped filtering:

The former keeps rows with ‘mass’ greater than the global average whereas the latter keeps rows with ‘mass’ greater than the gender average.

Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages:

See Also

[filter_all()], [filter_if()] and [filter_at()].

Other single table verbs: arrange(), mutate(), rename(), summarise()

Examples

data(se)

se |> tidybulk() |> filter(dex="untrt")

# Learn more in ?dplyr_tidy_eval

flybaseIDs

Description

flybaseIDs

Usage

flybaseIDs

Format

An object of class character of length 14599.
get_bibliography  

*Produces the bibliography list of your workflow*

**Description**

`get_bibliography()` takes as input a `tidybulk`

**Usage**

```r
get_bibliography(.data)
```

## S4 method for signature 'tbl'
```r
get_bibliography(.data)
```

## S4 method for signature 'tbl_df'
```r
get_bibliography(.data)
```

## S4 method for signature 'spec_tbl_df'
```r
get_bibliography(.data)
```

## S4 method for signature 'tidybulk'
```r
get_bibliography(.data)
```

## S4 method for signature 'SummarizedExperiment'
```r
get_bibliography(.data)
```

## S4 method for signature 'RangedSummarizedExperiment'
```r
get_bibliography(.data)
```

**Arguments**

- `.data`  
  A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to `tibble` with library(tidySummarizedExperiment))

**Details**

'r lifecycle::badge("maturing")'

This method returns the bibliography list of your workflow from the internals of a `tidybulk` object (attr(., "internals"))

**Value**

NULL. It prints a list of bibliography references for the software used through the workflow.

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).
A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

Examples

```r
get_bibliography(tidybulk::se_mini)
```

---

**get_reduced_dimensions_UMAP_bulk**  
*Get UMAP*

### Description

Get UMAP

### Usage

```r
get_reduced_dimensions_UMAP_bulk(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  calculate_for_pca_dimensions = min(20, top),
  ...
)
```

### Arguments

- `.data` A tibble
- `.element` A column symbol. The column that is used to calculate distance (i.e., normally samples)
- `.feature` A column symbol. The column that is represents entities to cluster (i.e., normally genes)
- `.abundance` A column symbol with the value the clustering is based on (e.g., ‘count’)
- `.dims` A integer vector corresponding to principal components of interest (e.g., 1:6)
- `top` An integer. How many top genes to select
- `of_samples` A boolean
get_reduced_dimensions_UMAP_bulk_SE

Calculate UMAP

Usage

get_reduced_dimensions_UMAP_bulk_SE(
  .data,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = NULL,
  calculate_for_pca_dimensions = min(20, top),
  ...
)

Arguments

.data A tibble
.dims A integer vector corresponding to principal components of interest (e.g., 1:6)
top An integer. How many top genes to select
of_samples A boolean
transform A function that will transform the counts, by default it is log1p for RNA sequencing data, but for avoiding transformation you can use identity
calculate_for_pca_dimensions An integer of length one. The number of PCA dimensions to based the UMAP calculation on. If NULL all variable features are considered
... Further parameters passed to the function uwot
.abundance A column symbol with the value the clustering is based on (e.g., ‘count’)

Value

A tibble with additional columns
**group_by**

A column symbol. The column that represents entities to cluster (i.e., normally genes)

A column symbol. The column that is used to calculate distance (i.e., normally samples)

**Value**

A tibble with additional columns

---

**group_by**

*Group by one or more variables*

**Description**

Most data operations are done on groups defined by variables. `group_by()` takes an existing tbl and converts it into a grouped tbl where operations are performed "by group". `ungroup()` removes grouping.

**Arguments**

- **.data**
  - A tbl. (See dplyr)
- **...**
  - In `group_by()`, variables or computations to group by. In `ungroup()`, variables to remove from the grouping.
- **.add**
  - When `FALSE`, the default, `group_by()` will override existing groups. To add to the existing groups, use `.add = TRUE`.
  - This argument was previously called `add`, but that prevented creating a new grouping variable called `add`, and conflicts with our naming conventions.
- **.drop**
  - When `.drop = TRUE`, empty groups are dropped. See [group_by_drop_default()] for what the default value is for this argument.

**Value**

A [grouped data frame][grouped_df()], unless the combination of `...` and `add` yields a non empty set of grouping columns, a regular (ungrouped) data frame otherwise.

**Methods**

These function are **generic**s, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

**Examples**

```r
by_cyl <- mtcars |> group_by(cyl)
```
Description

identify_abundant() takes as input a 'tbl' (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

Usage

identify_abundant(  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  factor_of_interest = NULL,
  minimum_counts = 10,
  minimum_proportion = 0.7)

## S4 method for signature 'spec_tbl_df'  identify_abundant(  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  factor_of_interest = NULL,
  minimum_counts = 10,
  minimum_proportion = 0.7)

## S4 method for signature 'tbl_df'  identify_abundant(  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  factor_of_interest = NULL,
  minimum_counts = 10,
  minimum_proportion = 0.7)

## S4 method for signature 'tidybulk'  identify_abundant(  .data,
identify_abundant

```r
identifi
```
impute_missing_abundance

**Details**

`r lifecycle::badge("maturing")`

At the moment this function uses edgeR (DOI: 10.1093/bioinformatics/btp616)

Underlying method: edgeR::filterByExpr( data, min.count = minimum_counts, group = string_factor_of_interest, min.prop = minimum_proportion )

**Value**

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A `SummarizedExperiment` object

A `SummarizedExperiment` object

**Examples**

```r
identify_abundant(
  tidybulk::se_mini
)
```

---

**impute_missing_abundance**

*impute transcript abundance if missing from sample-transcript pairs*

**Description**

`impute_missing_abundance()` takes as input A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional sample-transcript pairs with imputed transcript abundance.
Usage

```r
impute_missing_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  suffix = "",
  force_scaling = FALSE
)
```

## S4 method for signature 'spec_tbl_df'
impute_missing_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  suffix = "",
  force_scaling = FALSE
)

## S4 method for signature 'tbl_df'
impute_missing_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  suffix = "",
  force_scaling = FALSE
)

## S4 method for signature 'tidybulk'
impute_missing_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  suffix = "",
  force_scaling = FALSE
)

## S4 method for signature 'SummarizedExperiment'
impute_missing_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  suffix = "",
  force_scaling = FALSE
)
### Arguments

- `.data`  
  A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

- `.formula`  
  A formula with no response variable, representing the desired linear model where the first covariate is the factor of interest and the second covariate is the unwanted variation (of the kind `~ factor_of_interest + batch`)

- `.sample`  
  The name of the sample column

- `.transcript`  
  The name of the transcript/gene column

- `.abundance`  
  The name of the transcript/gene abundance column

- `.suffix`  
  A character string. This is added to the imputed count column names. If empty the count column are overwritten

- `.force_scaling`  
  A boolean. In case a abundance-containing column is not scaled (columns with `_scale` suffix), setting `force_scaling = TRUE` will result in a scaling by library size, to compensating for a possible difference in sequencing depth.

### Details

`r lifecycle::badge("maturing")`

This function imputes the abundance of missing sample-transcript pair using the median of the sample group defined by the formula

### Value

- A consistent object (to the input) non-sparse abundance
- A consistent object (to the input) with imputed abundance
- A consistent object (to the input) with imputed abundance
A consistent object (to the input) with imputed abundance
A 'SummarizedExperiment' object
A 'SummarizedExperiment' object

Examples

res =
impute_missing_abundance(
tidybulk::se_mini,
~ condition
)

Description

Inner join datasets
Right join datasets
Full join datasets

Arguments

x  
tbls to join. (See dplyr)
y  
tbls to join. (See dplyr)
by  
A character vector of variables to join by. (See dplyr)
copy  
If x and y are not from the same data source, and copy is TRUE, then y will be
copied into the same src as x. (See dplyr)
suffix  
If there are non-joined duplicate variables in x and y, these suffixes will be added
to the output to disambiguate them. Should be a character vector of length 2.
(See dplyr)
...  
Data frames to combine (See dplyr)

Value

A tt object
A tt object
A tt object
Examples

```r
annotation = tidybulk::se_mini |> tidybulk() |> as_tibble() |> distinct(.sample) |> mutate(source = "AU")
tidybulk::se_mini |> tidybulk() |> as_tibble() |> inner_join(annotation)

annotation = tidybulk::se_mini |> tidybulk() |> as_tibble() |> distinct(.sample) |> mutate(source = "AU")
tidybulk::se_mini |> tidybulk() |> as_tibble() |> right_join(annotation)

annotation = tidybulk::se_mini |> tidybulk() |> as_tibble() |> distinct(.sample) |> mutate(source = "AU")
tidybulk::se_mini |> tidybulk() |> as_tibble() |> full_join(annotation)
```

---

**keep_abundant**

---

**Keep abundant transcripts**

**Description**

`keep_abundant()` takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

**Usage**

```r
keep_abundant(
  .data, .sample = NULL, .transcript = NULL, .abundance = NULL, 
  factor_of_interest = NULL, 
  minimum_counts = 10, 
  minimum_proportion = 0.7
)
```

## S4 method for signature 'spec_tbl_df'
```
keep_abundant(
  .data, .sample = NULL, .transcript = NULL, 
  .abundance = NULL, 
  factor_of_interest = NULL, 
  minimum_counts = 10, 
  minimum_proportion = 0.7
)
```

## S4 method for signature 'tbl_df'

```
keep_abundant(
  .data, .sample = NULL, .transcript = NULL, 
  .abundance = NULL, 
  factor_of_interest = NULL, 
  minimum_counts = 10, 
  minimum_proportion = 0.7
)
```
keep_abundant(  .data,  .sample = NULL,  .transcript = NULL,  .abundance = NULL,  factor_of_interest = NULL,  minimum_counts = 10,  minimum_proportion = 0.7)

## S4 method for signature 'tidybulk'  keep_abundant(  .data,  .sample = NULL,  .transcript = NULL,  .abundance = NULL,  factor_of_interest = NULL,  minimum_counts = 10,  minimum_proportion = 0.7)

## S4 method for signature 'SummarizedExperiment'  keep_abundant(  .data,  .sample = NULL,  .transcript = NULL,  .abundance = NULL,  factor_of_interest = NULL,  minimum_counts = 10,  minimum_proportion = 0.7)

## S4 method for signature 'RangedSummarizedExperiment'  keep_abundant(  .data,  .sample = NULL,  .transcript = NULL,  .abundance = NULL,  factor_of_interest = NULL,  minimum_counts = 10,  minimum_proportion = 0.7)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
The name of the sample column

The name of the transcript/gene column

The name of the transcript/gene abundance column

The name of the column of the factor of interest. This is used for defining sample groups for the filtering process. It uses the filterByExpr function from edgeR.

A real positive number. It is the threshold of count per million that is used to filter transcripts/genes out from the scaling procedure.

A real positive number between 0 and 1. It is the threshold of proportion of samples for each transcripts/genes that have to be characterised by a cmp bigger than the threshold to be included for scaling procedure.

At the moment this function uses edgeR (DOI: 10.1093/bioinformatics/btp616)

Underlying method: edgeR::filterByExpr( data, min.count = minimum_counts, group = string_factor_of_interest, min.prop = minimum_proportion )

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A ‘SummarizedExperiment’ object

A ‘SummarizedExperiment’ object

keep_abundant(
  tidybulk::se_mini
)
**Description**

`keep_variable()` takes as input a 'tbl' (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

**Usage**

```r
default
keep_variable(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  top = 500,
  transform = log1p,
  log_transform = TRUE
)

## S4 method for signature 'spec_tbl_df'
default
keep_variable(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  top = 500,
  transform = log1p,
  log_transform = NULL
)

## S4 method for signature 'tbl_df'
default
keep_variable(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  top = 500,
  transform = log1p,
  log_transform = NULL
)

## S4 method for signature 'tidybulk'
default
keep_variable(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  top = 500,
  transform = log1p,
  log_transform = NULL
)
```
`keep_variable`

```r
.keep_variable = NULL,
.transform = NULL,
.abundance = NULL,
.top = 500,
.transform = log1p,
.log_transform = NULL
)
```

## S4 method for signature 'SummarizedExperiment'
`keep_variable(.data, top = 500, transform = log1p)`

## S4 method for signature 'RangedSummarizedExperiment'
`keep_variable(.data, top = 500, transform = log1p)`

### Arguments

- `.data` A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with `library(tidySummarizedExperiment)`
- `.sample` The name of the sample column
- `.transcript` The name of the transcript/gene column
- `.abundance` The name of the transcript/gene abundance column
- `.top` Integer. Number of top transcript to consider
- `.transform` A function that will transform the counts, by default it is log1p for RNA sequencing data, but for avoiding transformation you can use identity
- `.log_transform` DEPRECATED - A boolean, whether the value should be log-transformed (e.g., TRUE for RNA sequencing data)

### Details

`r lifecycle::badge("maturing")`

At the moment this function uses edgeR https://doi.org/10.1093/bioinformatics/btp616

### Value

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

Underlying method:  
s <- rowMeans(x - rowMeans(x)) ^ 2  
o <- order(s, decreasing = TRUE)  
gx <- x[o[1L:top] , drop = FALSE]  
variable_transcripts = rownames(x)

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).
A ‘SummarizedExperiment’ object

Examples

keep_variable(tidybulk::se_mini, top = 500)

---

**Description**

It perform log scaling and reverse the axis. Useful to plot negative log probabilities. To not be used directly but with ggplot (e.g. `scale_y_continuous(trans = "log10_reverse")`)

**Usage**

```
log10_reverse_trans()
```

**Details**

'r lifecycle::badge("maturing")'

**Value**

A scales object

**Examples**

```
library(ggplot2)
library(tibble)
tibble(pvalue = c(0.001, 0.05, 0.1), fold_change = 1:3) %>%
ggplot(aes(fold_change , pvalue)) +
  geom_point() +
  scale_y_continuous(trans = "log10_reverse")
```
logit_trans  logit scale

**Description**

it perform logit scaling with right axis formatting. To not be used directly but with ggplot (e.g. scale_y_continuous(trans = "log10_reverse") )

**Usage**

logit_trans()

**Details**

‘r lifecycle::badge("maturing")’

**Value**

A scales object

**Examples**

```r
library(ggplot2)
library(tibble)

tibble(pvalue = c(0.001, 0.05, 0.1), fold_change = 1:3) %>%
ggplot(aes(fold_change, pvalue)) +
geom_point() +
scale_y_continuous(trans = "log10_reverse")
```

---

**mutate**  Create, modify, and delete columns

**Description**

‘mutate()’ adds new variables and preserves existing ones; ‘transmute()’ adds new variables and drops existing ones. New variables overwrite existing variables of the same name. Variables can be removed by setting their value to ‘NULL’.
mutate

Arguments

.data A tbl. (See dplyr)

...<['tidy-eval'][dplyr_tidy_eval]> Name-value pairs. The name gives the name of
the column in the output.

The value can be:

* A vector of length 1, which will be recycled to the correct length. * A vector
the same length as the current group (or the whole data frame if ungrouped).
* ‘NULL’, to remove the column. * A data frame or tibble, to create multiple
columns in the output.

Value

An object of the same type as ‘.data’.

For ‘mutate()’:

* Rows are not affected. * Existing columns will be preserved unless explicitly modified. * New
columns will be added to the right of existing columns. * Columns given value ‘NULL’ will be
removed * Groups will be recomputed if a grouping variable is mutated. * Data frame attributes are
preserved.

For ‘transmute()’:

* Rows are not affected. * Apart from grouping variables, existing columns will be remove unless
explicitly kept. * Column order matches order of expressions. * Groups will be recomputed if a
grouping variable is mutated. * Data frame attributes are preserved.

Useful mutate functions

* `[+], [-], [log], etc., for their usual mathematical meanings
* [lead()], [lag()]
* [dense_rank()], [min_rank()], [percent_rank()], [row_number()], [cume_dist()], [ntile()]
* [cumsum()], [cummean()], [cummin()], [cummax()], [cumany()], [cumall()]
* [na_if()], [coalesce()]
* [if_else()], [recode()], [case_when()]

Grouped tibbles

Because mutating expressions are computed within groups, they may yield different results on
grouped tibbles. This will be the case as soon as an aggregating, lagging, or ranking function is
involved. Compare this ungrouped mutate:

With the grouped equivalent:

The former normalises ‘mass’ by the global average whereas the latter normalises by the averages
within gender levels.
Methods

These functions are **generic**s, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

Methods available in currently loaded packages:

See Also

Other single table verbs: `arrange()`, `filter()`, `rename()`, `summarise()`

Examples

# Newly created variables are available immediately
mtcars |> as_tibble() |> mutate(
  cyl2 = cyl * 2,
  cyl4 = cyl2 * 2
)

pivot_sample Extract sample-wise information

Description

`pivot_sample()` takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a ‘tbl’ with only sample-related columns

Usage

```
pivot_sample(.data, .sample = NULL)
## S4 method for signature 'spec_tbl_df'
pivot_sample(.data, .sample = NULL)
## S4 method for signature 'tbl_df'
pivot_sample(.data, .sample = NULL)
## S4 method for signature 'tidybulk'
pivot_sample(.data, .sample = NULL)
## S4 method for signature 'SummarizedExperiment'
pivot_sample(.data, .sample = NULL)
## S4 method for signature 'RangedSummarizedExperiment'
pivot_sample(.data, .sample = NULL)
```
pivot_transcript

Arguments

.data  A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.sample The name of the sample column

Details

‘r lifecycle::badge("maturing")’

This function extracts only sample-related information for downstream analysis (e.g., visualisation). It is disruptive in the sense that it cannot be passed anymore to tidybulk function.

Value

A ‘tbl’ with transcript-related information
A consistent object (to the input)
A consistent object (to the input)

Examples

pivot_sample(tidybulk::se_mini )

Description

pivot_transcript() takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a ‘tbl’ with only transcript-related columns

Usage

pivot_transcript(.data, .transcript = NULL)

## S4 method for signature 'spec_tbl_df'
pivot_transcript(.data, .transcript = NULL)

## S4 method for signature 'tbl_df'
pivot_transcript(.data, .transcript = NULL)

## S4 method for signature 'tidybulk'
pivot_transcript(.data, .transcript = NULL)
## S4 method for signature 'SummarizedExperiment'
pivot_transcript(.data, .transcript = NULL)

## S4 method for signature 'RangedSummarizedExperiment'
pivot_transcript(.data, .transcript = NULL)

### Arguments

.data
A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.transcript
The name of the transcript column

### Details

‘r lifecycle::badge("maturing")’

This function extracts only transcript-related information for downstream analysis (e.g., visualisation). It is disruptive in the sense that it cannot be passed anymore to tidybulk function.

### Value

A ‘tbl’ with transcript-related information
A consistent object (to the input)
A consistent object (to the input)

### Examples

pivot_transcript(tidybulk::se_mini )

---

**quantile_normalise_abundance**

Normalise by quantiles the counts of transcripts/genes

---

**Description**

quantile_normalise_abundance() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and scales transcript abundance compensating for sequencing depth (e.g., with TMM algorithm, Robinson and Oshlack doi.org/10.1186/gb-2010-11-3-r25).
Usage

quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = "add"
)

## S4 method for signature 'spec_tbl_df'
quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = "add"
)

## S4 method for signature 'tbl_df'
quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = "add"
)

## S4 method for signature 'tidybulk'
quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = "add"
)

## S4 method for signature 'SummarizedExperiment'
quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = NULL
)
## S4 method for signature 'RangedSummarizedExperiment'
quantile_normalise_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "limma_normalize_quantiles",
  action = NULL
)

### Arguments

- **.data**: A `tbl` (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
- **.sample**: The name of the sample column
- **.transcript**: The name of the transcript/gene column
- **.abundance**: The name of the transcript/gene abundance column
- **method**: A character string. Either "limma_normalize_quantiles" for limma::normalizeQuantiles or "preprocesscore_normalize_quantiles_use_target" for preprocessCore::normalize.quantiles.use.target for large-scale dataset, where limma could not be compatible.
- **action**: A character string between "add" (default) and "only". "add" joins the new information to the input tbl (default), "only" return a non-redundant tbl with the just new information.

### Details

'r lifecycle::badge("maturing")'

Scales transcript abundance compensating for sequencing depth (e.g., with TMM algorithm, Robinson and Oshlack doi.org/10.1186/gb-2010-11-3-r25). Lowly transcribed transcripts/genes (defined with minimum_counts and minimum_proportion parameters) are filtered out from the scaling procedure. The scaling inference is then applied back to all unfiltered data.

Underlying method edgeR::calcNormFactors(.data, method = c("TMM", "TMMwsp", "RLE","upperquartile"))

### Value

A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A 'SummarizedExperiment' object
A 'SummarizedExperiment' object
reduce_dimensions

Examples

```r
tidybulk::se_mini |> quantile_normalise_abundance()
```

---

**reduce_dimensions**  
*Dimension reduction of the transcript abundance data*

**Description**

reduce_dimensions() takes as input a `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and calculates the reduced dimensional space of the transcript abundance.

**Usage**

```r
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  action = "add",
  ...,
  log_transform = NULL
)
```

```r
## S4 method for signature 'spec_tbl_df'
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
```
reduce_dimensions

scale = TRUE,
action = "add",
...
log_transform = NULL
)

## S4 method for signature 'tbl_df'
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  action = "add",
  ...
  log_transform = NULL
)

## S4 method for signature 'tidybulk'
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  action = "add",
  ...
  log_transform = NULL
)

## S4 method for signature 'SummarizedExperiment'
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
reduce_dimensions

```r
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  action = "add",
  ...
  log_transform = NULL
)
```

## S4 method for signature 'RangedSummarizedExperiment'

```r
reduce_dimensions(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  .dims = 2,
  top = 500,
  of_samples = TRUE,
  transform = log1p,
  scale = TRUE,
  action = "add",
  ...
  log_transform = NULL
)
```

### Arguments

- **.data**: A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
- **.element**: The name of the element column (normally samples).
- **.feature**: The name of the feature column (normally transcripts/genes).
- **.abundance**: The name of the column including the numerical value the clustering is based on (normally transcript abundance).
- **method**: A character string. The dimension reduction algorithm to use (PCA, MDS, tSNE).
- **.dims**: An integer. The number of dimensions your are interested in (e.g., 4 for returning the first four principal components).
- **top**: An integer. How many top genes to select for dimensionality reduction.
- **of_samples**: A boolean. In case the input is a tidybulk object, it indicates Whether the element column will be sample or transcript column.
- **transform**: A function that will tranform the counts, by default it is log1p for RNA sequencing data, but for avoiding tranformation you can use identity.
- **scale**: A boolean for method="PCA", this will be passed to the 'prcomp' function. It is not included in the ... argument because although the default for 'prcomp' if FALSE, it is advisable to set it as TRUE.
**action**  A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).

... Further parameters passed to the function prcomp if you choose method="PCA" or Rtsne if you choose method="tSNE"

**log_transform**  DEPRECATED - A boolean, whether the value should be log-transformed (e.g., TRUE for RNA sequencing data)

**Details**

`r lifecycle::badge("maturing")`

This function reduces the dimensions of the transcript abundances. It can use multi-dimensional scaling (MDS; DOI.org/10.1186/gb-2010-11-3-r25), principal component analysis (PCA), or tSNE (Jesse Krijthe et al. 2018)

Underlying method for PCA: prcomp(scale = scale, ...)

Underlying method for MDS: limma::plotMDS(ndim = .dims, plot = FALSE, top = top)

Underlying method for tSNE: Rtsne::Rtsne(data, ...)

Underlying method for UMAP:

df_source = .data |>  
# Filter NA symbol filter(!.feature |> is.na() |> not()) |>  
# Prepare data frame distinct(!.feature,!!.element,!!.abundance) |>  
# Filter most variable genes keep_variable_transcripts(top) |> reduce_dimensions(method="PCA", .dims = calculate_for_pca_dimensions, action="get") |> as_matrix(rownames = quo_name(.element)) |> uwot::tumap(...)  

**Value**

A tbl object with additional columns for the reduced dimensions

A tbl object with additional columns for the reduced dimensions

A tbl object with additional columns for the reduced dimensions

A tbl object with additional columns for the reduced dimensions

A ‘SummarizedExperiment’ object

A ‘SummarizedExperiment’ object

**Examples**

```r
counts.MDS = tidybulk::se_mini |> identify_abundant() |> reduce_dimensions( method="MDS", .dims = 3)
```

```r
counts.PCA = tidybulk::se_mini |
```
identify_abundant() |> 
reduce_dimensions(method="PCA", .dims = 3)

---

**reexports**

*Objects exported from other packages*

**Description**

These objects are imported from other packages. Follow the links below to see their documentation.

- **dplyr** `do, select`
- **tibble** `as_tibble, tibble`

---

**remove_redundancy**

*Drop redundant elements (e.g., samples) for which feature (e.g., transcript/gene) abundances are correlated*

**Description**

`remove_redundancy()` takes as input a `tbl` (with at least three columns for sample, feature and transcript abundance) or a `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) for correlation method or `| <DIMENSION 1> | <DIMENSION 2> | ... |` for reduced_dimensions method, and returns a consistent object (to the input) with dropped elements (e.g., samples).

**Usage**

```r
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column,
  Dim_b_column,
  log_transform = NULL
)
```
## S4 method for signature 'spec_tbl_df'
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column = NULL,
  Dim_b_column = NULL,
  log_transform = NULL
)

## S4 method for signature 'tbl_df'
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column = NULL,
  Dim_b_column = NULL,
  log_transform = NULL
)

## S4 method for signature 'tidybulk'
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column = NULL,
  Dim_b_column = NULL,
  log_transform = NULL
)
remove_redundancy

## S4 method for signature 'SummarizedExperiment'
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column = NULL,
  Dim_b_column = NULL,
  log_transform = NULL
)

## S4 method for signature 'RangedSummarizedExperiment'
remove_redundancy(
  .data,
  .element = NULL,
  .feature = NULL,
  .abundance = NULL,
  method,
  of_samples = TRUE,
  correlation_threshold = 0.9,
  top = Inf,
  transform = identity,
  Dim_a_column = NULL,
  Dim_b_column = NULL,
  log_transform = NULL
)

### Arguments

- **.data**
  - A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

- **.element**
  - The name of the element column (normally samples).

- **.feature**
  - The name of the feature column (normally transcripts/genes).

- **.abundance**
  - The name of the column including the numerical value the clustering is based on (normally transcript abundance).

- **method**
  - A character string. The method to use, correlation and reduced_dimensions are available. The latter eliminates one of the most proximar pairs of samples in PCA reduced dimensions.

- **of_samples**
  - A boolean. In case the input is a tidybulk object, it indicates Whether the element column will be sample or transcript column.

- **correlation_threshold**
  - A real number between 0 and 1. For correlation based calculation.
remove_redundancy

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>top</td>
<td>An integer. How many top genes to select for correlation based method</td>
</tr>
<tr>
<td>transform</td>
<td>A function that will transform the counts, by default it is log1p for RNA sequencing data, but for avoiding transformation you can use identity</td>
</tr>
<tr>
<td>Dim_a_column</td>
<td>A character string. For reduced_dimension based calculation. The column of one principal component</td>
</tr>
<tr>
<td>Dim_b_column</td>
<td>A character string. For reduced_dimension based calculation. The column of another principal component</td>
</tr>
<tr>
<td>log_transform</td>
<td>DEPRECATED - A boolean, whether the value should be log-transformed (e.g., TRUE for RNA sequencing data)</td>
</tr>
</tbody>
</table>

Details

'r lifecycle::badge("maturing")'

This function removes redundant elements from the original data set (e.g., samples or transcripts). For example, if we want to define cell-type specific signatures with low sample redundancy. This function returns a tibble with dropped redundant elements (e.g., samples). Two redundancy estimation approaches are supported: (i) removal of highly correlated clusters of elements (keeping a representative) with method="correlation"; (ii) removal of most proximal element pairs in a reduced dimensional space.

Underlying method for correlation: widyr::pairwise_cor(sample, transcript,count, sort = TRUE, diag = FALSE, upper = FALSE)

Underlying custom method for reduced dimensions: select_closest_pairs = function(df) couples <-

while (df |> nrow() > 0) pair <- df |> arrange(dist) |> head(n = 1) couples <- couples |> bind_rows(pair)

df <- df |> filter(¡sample 1' ¡sample 2' )

couples

Value

A tbl object with with dropped redundant elements (e.g., samples).

A tbl object with with dropped redundant elements (e.g., samples).

A tbl object with with dropped redundant elements (e.g., samples).

A tbl object with with dropped redundant elements (e.g., samples).

A 'SummarizedExperiment' object

A 'SummarizedExperiment' object

Examples

```r
tidybulk::se_mini |> identify_abundant() |> remove_redundancy(
  .element = sample,
  .feature = transcript,
  .abundance = count,
  ```
counts.MDS =
tidybulk::se_mini |>
identify_abundant() |>
  reduce_dimensions( method="MDS", .dims = 3)

remove_redundancy(
counts.MDS,
Dim_a_column = "Dim1",
Dim_b_column = "Dim2",
.element = sample,
  method = "reduced_dimensions"
)

---

### rename

**Rename columns**

#### Description

Rename individual variables using `new_name = old_name` syntax.

#### Arguments

- `.data`
  
  A tbl. (See dplyr)

- `...`
  
  Use `new_name = old_name` to rename selected variables.

#### Value

An object of the same type as `.data`. * Rows are not affected. * Column names are changed; column order is preserved * Data frame attributes are preserved. * Groups are updated to reflect new names.

#### Scoped selection and renaming

Use the three scoped variants ([rename_all()](#)), [rename_if()](#), [rename_at()](#) to renaming a set of variables with a function.

#### Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages:
See Also

Other single table verbs: arrange(), filter(), mutate(), summarise()
## S4 method for signature 'tidybulk'
rotate_dimensions(
  .data,
  dimension_1_column,
  dimension_2_column,
  rotation_degrees,
  .element = NULL,
  of_samples = TRUE,
  dimension_1_column_rotated = NULL,
  dimension_2_column_rotated = NULL,
  action = "add"
)

## S4 method for signature 'SummarizedExperiment'
rotate_dimensions(
  .data,
  dimension_1_column,
  dimension_2_column,
  rotation_degrees,
  .element = NULL,
  of_samples = TRUE,
  dimension_1_column_rotated = NULL,
  dimension_2_column_rotated = NULL,
  action = "add"
)

## S4 method for signature 'RangedSummarizedExperiment'
rotate_dimensions(
  .data,
  dimension_1_column,
  dimension_2_column,
  rotation_degrees,
  .element = NULL,
  of_samples = TRUE,
  dimension_1_column_rotated = NULL,
  dimension_2_column_rotated = NULL,
  action = "add"
Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

dimension_1_column

A character string. The column of the dimension 1

dimension_2_column

A character string. The column of the dimension 2

rotation_degrees

A real number between 0 and 360

.element

The name of the element column (normally samples).

of_samples

A boolean. In case the input is a tidybulk object, it indicates Whether the element column will be sample or transcript column

dimension_1_column_rotated

A character string. The column of the rotated dimension 1 (optional)

dimension_2_column_rotated

A character string. The column of the rotated dimension 2 (optional)

action

A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).

Details

‘r lifecycle::badge("maturing")’

This function to rotate two dimensions such as the reduced dimensions.

Underlying custom method: rotation = function(m, d) // r = the angle // m data matrix r = d * pi / 180 ((bind_rows( c(‘1’ = cos(r), ‘2’ = -sin(r)), c(‘1’ = sin(r), ‘2’ = cos(r)) ) |> as_matrix())

Value

A tbl object with additional columns for the reduced dimensions. additional columns for the rotated dimensions. The rotated dimensions will be added to the original data set as ‘<NAME OF DIMENSION> rotated <ANGLE>’ by default, or as specified in the input arguments.

A tbl object with additional columns for the reduced dimensions. additional columns for the rotated dimensions. The rotated dimensions will be added to the original data set as ‘<NAME OF DIMENSION> rotated <ANGLE>’ by default, or as specified in the input arguments.

A tbl object with additional columns for the reduced dimensions. additional columns for the rotated dimensions. The rotated dimensions will be added to the original data set as ‘<NAME OF DIMENSION> rotated <ANGLE>’ by default, or as specified in the input arguments.

A tbl object with additional columns for the reduced dimensions. additional columns for the rotated dimensions. The rotated dimensions will be added to the original data set as ‘<NAME OF DIMENSION> rotated <ANGLE>’ by default, or as specified in the input arguments.

A ‘SummarizedExperiment’ object

A ‘SummarizedExperiment’ object
Examples

counts.MDS =
  tidybulk::se_mini |>
  identify_abundant() |>
  reduce_dimensions( method="MDS", .dims = 3)

counts.MDS.rotated = rotate_dimensions(counts.MDS, `Dim1`, `Dim2`, rotation_degrees = 45, .element = sample)

Description

See [this repository](https://github.com/jennybc/row-oriented-workflows) for alternative ways to perform row-wise operations.

Arguments

data
  Input data frame.

... Variables to be preserved when calling summarise(). This is typically a set of variables whose combination uniquely identify each row. NB: unlike group_by() you can not create new variables here but instead you can select multiple variables with (e.g.) everything().

Details

‘rowwise()’ is used for the results of [do()] when you create list-variables. It is also useful to support arbitrary complex operations that need to be applied to each row.

Currently, rowwise grouping only works with data frames. Its main impact is to allow you to work with list-variables in [summarise()] and [mutate()] without having to use [[1]]. This makes ‘summarise()’ on a rowwise tbl effectively equivalent to [plyr::ldply()].

Value

A consistent object (to the input)
A ‘tbl’

Examples

df <- expand.grid(x = 1:3, y = 3:1)
df_done <- df |> rowwise()
scale_abundance

Scale the counts of transcripts/genes

Description

scale_abundance() takes as input a `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and scales transcript abundance compensating for sequencing depth (e.g., with TMM algorithm, Robinson and Oshlack doi.org/10.1186/gb-2010-11-3-r25).

Usage

```r
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = "add",
  reference_selection_function = NULL
)
```

```r
# S4 method for signature 'spec_tbl_df'
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = "add",
  reference_selection_function = NULL
)
```

```r
# S4 method for signature 'tbl_df'
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = "add",
)
```
scale_abundance

    reference_selection_function = NULL
  )

## S4 method for signature 'tidybulk'
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = "add",
  reference_selection_function = NULL
)

## S4 method for signature 'SummarizedExperiment'
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = NULL,
  reference_selection_function = NULL
)

## S4 method for signature 'RangedSummarizedExperiment'
scale_abundance(
  .data,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "TMM",
  reference_sample = NULL,
  .subset_for_scaling = NULL,
  action = NULL,
  reference_selection_function = NULL
)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.sample The name of the sample column
scale_abundance

- **.transcript**: The name of the transcript/gene column
- **.abundance**: The name of the transcript/gene abundance column
- **method**: A character string. The scaling method passed to the back-end function (i.e., edgeR::calcNormFactors; "TMM","TMMwsp","RLE","upperquartile")
- **reference_sample**: A character string. The name of the reference sample. If NULL the sample with highest total read count will be selected as reference.
- **.subset_for_scaling**: A gene-wise quosure condition. This will be used to filter rows (features/genes) of the dataset. For example
- **action**: A character string between "add" (default) and "only". "add" joins the new information to the input tbl (default), "only" return a non-redundant tbl with the just new information.
- **reference_selection_function**: DEPRECATED. please use reference_sample.

**Details**

'r lifecycle::badge("maturing")'

Scales transcript abundance compensating for sequencing depth (e.g., with TMM algorithm, Robinson and Oshlack doi.org/10.1186/gb-2010-11-3-r25). Lowly transcribed transcripts/genes (defined with minimum_counts and minimum_proportion parameters) are filtered out from the scaling procedure. The scaling inference is then applied back to all unfiltered data.

Underlying method edgeR::calcNormFactors(.data, method = c("TMM","TMMwsp","RLE","upperquartile"))

**Value**

A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A tbl object with additional columns with scaled data as `<NAME OF COUNT COLUMN>_scaled`
A `SummarizedExperiment` object
A `SummarizedExperiment` object

**Examples**

```r
tidybulk::se_mini |>
   identify_abundant() |>
   scale_abundance()
```
se

Description
SummarizedExperiment

Usage
se

Format
An object of class RangedSummarizedExperiment with 100 rows and 8 columns.

se_mini

Description
SummarizedExperiment mini for vignette

Usage
se_mini

Format
An object of class SummarizedExperiment with 527 rows and 5 columns.

summarise

Description
’summarise()’ creates a new data frame. It will have one (or more) rows for each combination of grouping variables; if there are no grouping variables, the output will have a single row summarising all observations in the input. It will contain one column for each grouping variable and one column for each of the summary statistics that you have specified.

‘summarise()’ and ‘summarize()’ are synonyms.
Arguments

.data  A tbl. (See dplyr)

...  <['tidy-eval']>[dplyr_tidy_eval]> Name-value pairs of summary functions. The name will be the name of the variable in the result. The value can be:
- A vector of length 1, e.g. ‘min(x)’, ‘n()’, or ‘sum(is.na(y))’. * A vector of length ‘n’, e.g. ‘quantile()’. * A data frame, to add multiple columns from a single expression.

Value

An object _usually_ of the same type as `.data`.

- The rows come from the underlying `group_keys()`. * The columns are a combination of the grouping keys and the summary expressions that you provide. * If `x` is grouped by more than one variable, the output will be another [grouped_df] with the right-most group removed. * If `x` is grouped by one variable, or is not grouped, the output will be a [tibble]. * Data frame attributes are **not** preserved, because `summarise()` fundamentally creates a new data frame.

Useful functions

* Center: [mean()], [median()] * Spread: [sd()], [IQR()], [mad()] * Range: [min()], [max()], [quantile()] * Position: [first()], [last()], [nth()] * Count: [n()], [n_distinct()] * Logical: [any()], [all()]

Backend variations

The data frame backend supports creating a variable and using it in the same summary. This means that previously created summary variables can be further transformed or combined within the summary, as in [mutate()]. However, it also means that summary variables with the same names as previous variables overwrite them, making those variables unavailable to later summary variables.

This behaviour may not be supported in other backends. To avoid unexpected results, consider using new names for your summary variables, especially when creating multiple summaries.

Methods

This function is a **generic**, which means that packages can provide implementations (methods) for other classes. See the documentation of individual methods for extra arguments and differences in behaviour.

The following methods are currently available in loaded packages:

See Also

Other single table verbs: arrange(), filter(), mutate(), rename()
symbol_to_entrez

Examples

# A summary applied to ungrouped tbl returns a single row

mtcars |>
  summarise(mean = mean(disp))

symbol_to_entrez  Get ENTREZ id from gene SYMBOL

Description

Get ENTREZ id from gene SYMBOL

Usage

symbol_to_entrez(.data, .transcript = NULL, .sample = NULL)

Arguments

.data  A tt or tbl object.
.transcript  A character. The name of the gene symbol column.
.sample  The name of the sample column

Value

A tbl

Examples

# This function was designed for data.frame
# Convert from SummarizedExperiment for this example. It is NOT recommended.

tidybulk::se_mini |> tidybulk() |> as_tibble() |> symbol_to_entrez(.transcript = .feature, .sample = .sample)
test_differential_abundance

Perform differential transcription testing using edgeR quasi-likelihood (QLT), edgeR likelihood-ratio (LR), limma-voom, limma-voom-with-quality-weights or DESeq2

Description

test_differential_abundance() takes as input A 'tbl' (with at least three columns for sample, feature and transcript abundance) or 'SummarizedExperiment' (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

Usage

test_differential_abundance(
    .data,
    .formula,
    .sample = NULL,
    .transcript = NULL,
    .abundance = NULL,
    contrasts = NULL,
    method = "edgeR_quasi_likelihood",
    test_above_log2_fold_change = NULL,
    scaling_method = "TMM",
    omit_contrast_in_colnames = FALSE,
    prefix = "",
    action = "add",
    ...
)

# S4 method for signature 'spec_tbl_df'
test_differential_abundance(
    .data,
    .formula,
    .sample = NULL,
    .transcript = NULL,
    .abundance = NULL,
    contrasts = NULL,
    method = "edgeR_quasi_likelihood",
    test_above_log2_fold_change = NULL,
    scaling_method = "TMM",
    omit_contrast_in_colnames = FALSE,
    prefix = "",
)
test_differential_abundance

action = "add",
...,  
significance_threshold = NULL,
fill_missing_values = NULL,
.contrasts = NULL  
)

## S4 method for signature 'tbl_df'

R.test_differential_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  contrasts = NULL,
  method = "edgeR_quasi_likelihood",
  test_above_log2_fold_change = NULL,
  scaling_method = "TMM",
  omit_contrast_in_colnames = FALSE,
  prefix = ",",
  action = "add",
  ...,  
significance_threshold = NULL,
fill_missing_values = NULL,
.contrasts = NULL
)

## S4 method for signature 'tidybulk'

R.test_differential_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  contrasts = NULL,
  method = "edgeR_quasi_likelihood",
  test_above_log2_fold_change = NULL,
  scaling_method = "TMM",
  omit_contrast_in_colnames = FALSE,
  prefix = ",",
  action = "add",
  ...,  
significance_threshold = NULL,
fill_missing_values = NULL,
.contrasts = NULL
)

## S4 method for signature 'SummarizedExperiment'
test_differential_abundance(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  contrasts = NULL,
  method = "edgeR_quasi_likelihood",
  test_above_log2_fold_change = NULL,
  scaling_method = "TMM",
  omit_contrast_in_colnames = FALSE,
  prefix = "",
  action = "add",
  ...
)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.formula A formula representing the desired linear model. If there is more than one factor, they should be in the order factor of interest + additional factors.

.sample The name of the sample column

.transcript The name of the transcript/gene column
.abundance  The name of the transcript/gene abundance column
contrasts  This parameter takes the format of the contrast parameter of the method of choice. For edgeR and limma-voom is a character vector. For DESeq2 is a list including a character vector of length three. The first covariate is the one the model is tested against (e.g., ~ factor_of_interest)
method  A string character. Either "edgeR_quasi_likelihood" (i.e., QLF), "edgeR_likelihood_ratio" (i.e., LRT), "edgeR_robust_likelihood_ratio", "DESeq2", "limma_voom", "limma_voom_sample_weights"
test_above_log2_fold_change  A positive real value. This works for edgeR and limma_voom methods. It uses the ‘treat’ function, which tests that the difference in abundance is bigger than this threshold rather than zero https://pubmed.ncbi.nlm.nih.gov/19176553.
scaling_method  A character string. The scaling method passed to the back-end functions: edgeR and limma-voom (i.e., edgeR::calcNormFactors; "TMM","TMMwsp","RLE","upperquartile"). Setting the parameter to "none" will skip the compensation for sequencing-depth for the method edgeR or limma-voom.
omit_contrast_in_colnames  If just one contrast is specified you can choose to omit the contrast label in the colnames.
prefix  A character string. The prefix you would like to add to the result columns. It is useful if you want to compare several methods.
action  A character string. Whether to join the new information to the input tbl (add), or just get the non-redundant tbl with the new information (get).
...  Further arguments passed to some of the internal functions. Currently, it is needed just for internal debug.
significance_threshold  DEPRECATED - A real between 0 and 1 (usually 0.05).
fill_missing_values  DEPRECATED - A boolean. Whether to fill missing sample/transcript values with the median of the transcript. This is rarely needed.
.contrasts  DEPRECATED - This parameter takes the format of the contrast parameter of the method of choice. For edgeR and limma-voom is a character vector. For DESeq2 is a list including a character vector of length three. The first covariate is the one the model is tested against (e.g., ~ factor_of_interest)

Details
‘r lifecycle::badge("maturing")’
This function provides the option to use edgeR https://doi.org/10.1093/bioinformatics/btp616, limma-voom https://doi.org/10.1186/gb-2014-15-2-r29, limma_voom_sample_weights https://doi.org/10.1093/nar/gkv412 or DESeq2 https://doi.org/10.1186/s13059-014-0550-8 to perform the testing. All methods use raw counts, irrespective of if scale_abundance or adjust_abundance have been calculated, therefore it is essential to add covariates such as batch effects (if applicable) in the formula.
Underlying method for edgeR framework: .data >
# Filter keep_abundant( factor_of_interest = !(as.symbol(parse_formula(.formula)[1])), minimum_counts = minimum_counts, minimum_proportion = minimum_proportion ) |> # Format select(!!.transcript,!!.sample,!!.abundance) |> spread(!!.sample,!!.abundance) |> as_matrix(rownames = !!.transcript) # edgeR edgeR::DGEList(counts = .) |> edgeR::calcNormFactors(method = scaling_method) |> edgeR::estimateDisp(design) |> # Fit edgeR::glmQLFit(design) |> // or glmFit according to choice edgeR::glmQLFTest(coef = 2, contrast = my_contrasts) // or glmLRT according to choice Underlying method for DESeq2 framework: keep_abundant( factor_of_interest = !(as.symbol(parse_formula(.formula)[1])), minimum_counts = minimum_counts, minimum_proportion = minimum_proportion ) |> # DESeq2 DESeq2::DESeqDataSet(design = .formula) |> DESeq2::DESeq() |> DESeq2::results() ---

**Value**

A consistent object (to the input) with additional columns for the statistics from the test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A ‘SummarizedExperiment’ object

A ‘SummarizedExperiment’ object

---

**Examples**

# edgeR

tidybulk::se_mini |> identify_abundant() |> test_differential_abundance( ~ condition )

# The function `test_differential_abundance` operates with contrasts too

tidybulk::se_mini |> identify_abundant(factor_of_interest = condition) |> test_differential_abundance( ~ 0 + condition, contrasts = c("conditionTRUE - conditionFALSE")

# DESeq2 - equivalent for limma-voom

my_se_mini = tidybulk::se_mini
my_se_mini$condition = factor(my_se_mini$condition)
# demonstrating with `fitType` that you can access any arguments to DESeq()
my_se_mini |>
   identify_abundant(factor_of_interest = condition) |>
   test_differential_abundance( - condition, method="deseq2", fitType="local")

# testing above a log2 threshold, passes along value to lfcThreshold of results()
res <- my_se_mini |>
   identify_abundant(factor_of_interest = condition) |>
   test_differential_abundance( - condition, method="deseq2",
                             fitType="local",
                             test_above_log2_fold_change=4 )

# Use random intercept and random effect models - NOT RUN BECAUSE MATRIX PACKAGE HAS A TEMPORARY BUG WITH LLME4
## Not run:
se_mini[1:50,] |>
   identify_abundant(factor_of_interest = condition) |>
   test_differential_abundance( ~ condition + (1 + condition | time),
                             method = "glmmseq_lme4", cores = 1
)
## End(Not run)

# confirm that lfcThreshold was used
## Not run:
res |> mcols() |> 
    DESeq2::DESeqResults() |> 
    DESeq2::plotMA()
## End(Not run)

# The function `test_differential_abundance` operates with contrasts too
my_se_mini |>
   identify_abundant() |>
   test_differential_abundance( ~ 0 + condition,
                              contrasts = list(c("condition", "TRUE", "FALSE")),
                              method="deseq2",
                              fitType="local"
)
test_differential_cellularity() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

Usage

test_differential_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  significance_threshold = 0.05,
  ...
)

## S4 method for signature 'spec_tbl_df'
test_differential_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  significance_threshold = 0.05,
  ...
)

## S4 method for signature 'tbl_df'
test_differential_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  significance_threshold = 0.05,
  ...
)

## S4 method for signature 'tidybulk'
test_differential_cellularity(

test_differential_cellularity

```r
.data, .formula, .sample = NULL, .transcript = NULL, .abundance = NULL,
method = "cibersort", reference = X_cibersort,
significance_threshold = 0.05,
...
```

## S4 method for signature 'SummarizedExperiment'

```r
test_differential_cellularity(
.data,
.formula,
.sample = NULL, .transcript = NULL, .abundance = NULL,
method = "cibersort", reference = X_cibersort,
significance_threshold = 0.05,
...
)
```

## S4 method for signature 'RangedSummarizedExperiment'

```r
test_differential_cellularity(
.data,
.formula,
.sample = NULL, .transcript = NULL, .abundance = NULL,
method = "cibersort", reference = X_cibersort,
significance_threshold = 0.05,
...
)
```

### Arguments

- `.data` A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

- `.formula` A formula representing the desired linear model. The formula can be of two forms: multivariable (recommended) or univariable Respectively: \"factor_of_interest ~ \." or \". ~ factor_of_interest\". The dot represents cell-type proportions, and it is mandatory. If censored regression is desired (coxph) the formula should be of the form \"survival::Surv(y, dead) ~ \".

- `.sample` The name of the sample column
.transcript  The name of the transcript/gene column
.abundance  The name of the transcript/gene abundance column
.method  A string character. Either \"cibersort\", \"epic\" or \"llsr\". The regression method will be chosen based on being multivariable: lm or cox-regression (both on logit-transformed proportions); or univariable: beta or cox-regression (on logit-transformed proportions). See .formula for multi- or univariable choice.
.reference  A data frame. The transcript/cell_type data frame of integer transcript abundance
.significance_threshold  A real between 0 and 1 (usually 0.05).
...  Further parameters passed to the method deconvolve_cellularity

Details

\'r lifecycle::badge("maturing")\'

This routine applies a deconvolution method (e.g., Cibersort; DOI: 10.1038/nmeth.3337) and passes the proportions inferred into a generalised linear model (DOI:dx.doi.org/10.1007/s11749-010-0189-z) or a cox regression model (ISBN: 978-1-4757-3294-8)

Underlying method for the generalised linear model: data |> deconvolve_cellularity( !!.sample, !!.transcript, !!.abundance, method=method, reference = reference, action="get", ... ) [...] betareg::betareg(.my_formula, .)

Underlying method for the cox regression: data |> deconvolve_cellularity( !!.sample, !!.transcript, !!.abundance, method=method, reference = reference, action="get", ... ) [...] mutate(.proportion_0_corrected |> boot::logit()) survival::coxph(.my_formula, .)

Value

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A `SummarizedExperiment` object

A `SummarizedExperiment` object

Examples

# Regular regression
test_differential_cellularity(
tidybulk::se_mini ,
  . ~ condition,
  cores = 1
)

# Cox regression - multiple
tidybulk::se_mini |> 

# Test
test_differential_cellularity(
test_gene_enrichment

survival::Surv(days, dead) ~ .,
cores = 1
)

test_gene_enrichment  analyse gene enrichment with EGSEA

Description

test_gene_enrichment() takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a ‘tbl’ of gene set information

Usage

test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
  method = NULL,
  .contrasts = NULL
)

## S4 method for signature 'spec_tbl_df'
test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
method = NULL,
    .contrasts = NULL
  )

## S4 method for signature 'tbl_df'

test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
  method = NULL,
  .contrasts = NULL
)

## S4 method for signature 'tidybulk'

test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
  method = NULL,
  .contrasts = NULL
)

## S4 method for signature 'SummarizedExperiment'

test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
test_gene_enrichment

```r
  "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
  method = NULL,
  .contrasts = NULL
)
```

```r
## S4 method for signature 'RangedSummarizedExperiment'
test_gene_enrichment(
  .data,
  .formula,
  .sample = NULL,
  .entrez,
  .abundance = NULL,
  contrasts = NULL,
  methods = c("camera", "roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease",
                "kegg_metabolism", "kegg_signaling"),
  species,
  cores = 10,
  method = NULL,
  .contrasts = NULL
)
```

**Arguments**

- `.data` A `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
- `.formula` A formula with no response variable, representing the desired linear model
- `.sample` The name of the sample column
- `.entrez` The ENTREZ ID of the transcripts/genes
- `.abundance` The name of the transcript/gene abundance column
- `.contrasts` This parameter takes the format of the contrast parameter of the method of choice. For edgeR and limma-voom is a character vector. For DESeq2 is a list including a character vector of length three. The first covariate is the one the model is tested against (e.g., ~ factor_of_interest)
- `.methods` A character vector. One or 3 or more methods to use in the testing (currently EGSEA errors if 2 are used). Type EGSEA::egsea.base() to see the supported GSE methods.
- `.gene_sets` A character vector or a list. It can take one or more of the following built-in collections as a character vector: c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease", "kegg_metabolism", "kegg_signaling"), to be used with EGSEA buildIdx. c1 is human specific. Alternatively, a list of user-supplied gene sets can be provided, to be used with EGSEA buildCustomIdx. In that case, each gene set is a character vector of Entrez IDs and the names of the list are the gene set names.
species  A character. It can be human, mouse or rat.
cores   An integer. The number of cores available
method  DEPRECATED. Please use methods.
.contrasts  DEPRECATED - This parameter takes the format of the contrast parameter of
the method of choice. For edgeR and limma-voom is a character vector. For
DESeq2 is a list including a character vector of length three. The first covariate
is the one the model is tested against (e.g., ~ factor_of_interest)

Details

\texttt{r lifecycle::badge("maturing")}'

This wrapper executes ensemble gene enrichment analyses of the dataset using EGSEA (DOI:0.12688/f1000research.12544.1)

dge = data |> keep_abundant( factor_of_interest = !!as.symbol(parse_formula(.formula)[[1]]), !!.sample, !!.entrez, !!.abundance )

# Make sure transcript names are adjacent [...] as_matrix(rownames = !!.entrez) edgeR::DGEList(counts = .)
idx = buildIdx(entrezIDs = rownames(dge), species = species, msigdb.gsets = msigdb.gsets, kegg.exclude = kegg.exclude)

dge |> # Calculate weights limma::voom(design, plot = FALSE) |> # Execute EGSEA egsea( contrasts = my_contrasts, baseGSEAs = methods, gs.annots = idx, sort.by = "med.rank", num.threads = cores, report = FALSE )

Value

A consistent object (to the input)
A consistent object (to the input)
A consistent object (to the input)
A consistent object (to the input)
A consistent object (to the input)
A consistent object (to the input)

Examples

## Not run:

library(SummarizedExperiment)
se = tidybulk::se_mini
rowData( se)$entrez = rownames(se)
df_entrez = aggregate_duplicates(se,.transcript = entrez)

library("EGSEA")

test_gene_enrichment(
  df_entrez,
test_gene_overrepresentation

- condition,
  .sample = sample,
  .entrez = entrez,
  .abundance = count,
  methods = c("roast", "safe", "gage", "padog", "globaltest", "ora"),
  gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease", "kegg_metabolism", "kegg_signal",
  species="human",
  cores = 2
)

## End(Not run)

test_gene_overrepresentation

*analyse gene over-representation with GSEA*

**Description**

test_gene_overrepresentation() takes as input a `tbl` (with at least three columns for sample, feature and transcript abundance) or `SummarizedExperiment` (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a ‘tbl’ with the GSEA statistics

**Usage**

test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

## S4 method for signature 'spec_tbl_df'

test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

## S4 method for signature 'tbl_df'

test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

## S4 method for signature 'tidybulk'
test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

## S4 method for signature 'SummarizedExperiment'
test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

## S4 method for signature 'RangedSummarizedExperiment'
test_gene_overrepresentation(
  .data,
  .entrez,
  .do_test,
  species,
  .sample = NULL,
  gene_sets = NULL,
  gene_set = NULL
)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))
### test_gene_overrepresentation

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<td>.entrez</td>
<td>The ENTREZ ID of the transcripts/genes</td>
</tr>
<tr>
<td>.do_test</td>
<td>A boolean column name symbol. It indicates the transcript to check</td>
</tr>
<tr>
<td>species</td>
<td>A character. For example, human or mouse. MSigDB uses the latin species names (e.g., &quot;Mus musculus&quot;, &quot;Homo sapiens&quot;)</td>
</tr>
<tr>
<td>.sample</td>
<td>The name of the sample column</td>
</tr>
<tr>
<td>gene_sets</td>
<td>A character vector. The subset of MSigDB datasets you want to test against (e.g. &quot;C2&quot;). If NULL all gene sets are used (suggested). This argument was added to avoid time overflow of the examples.</td>
</tr>
</tbody>
</table>

#### Details

'\r lifecycle::badge("maturing")'

This wrapper executes gene enrichment analyses of the dataset using a list of transcripts and GSEA. This wrapper uses clusterProfiler (DOI: doi.org/10.1089/omi.2011.0118) on the back-end.

Underlying method: `msigdb::msigdb(species = species) |> nest(data = -gs_cat) |> mutate(test = map(data, ~ clusterProfiler::enricher(my_entrez_rank, TERM2GENE=.x |> select(gs_name, entrez_gene), pvalueCutoff = 1)) |> as_tibble())`

#### Value

A consistent object (to the input)

- A `spec_tbl_df` object
- A `tbl_df` object
- A `tidybulk` object
- A `SummarizedExperiment` object
- A `RangedSummarizedExperiment` object

#### Examples

```r
print("Not run for build time.")

# se_mini = aggregate_duplicates(tidybulk::se_mini, .transcript = entrez)
# df_entrez = mutate(df_entrez, do_test = feature %in% c("TNFRSF4", "PLCH2", "PADI4", "PAX7"))

## Not run:
 test_gene_overrepresentation(
  df_entrez,
  .sample = sample,
  .entrez = entrez,
  .do_test = do_test,
  species="Homo sapiens",
  gene_sets =c("C2")
 )

## End(Not run)
```
Description

test_gene_rank() takes as input a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a ‘tbl’ with the GSEA statistics

Usage

test_gene_rank(
    .data,
    .entrez,
    .arrange_desc,
    species,
    .sample = NULL,
    gene_sets = NULL,
    gene_set = NULL
)

## S4 method for signature 'spec_tbl_df'
test_gene_rank(
    .data,
    .entrez,
    .arrange_desc,
    species,
    .sample = NULL,
    gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7"),
    gene_set = NULL
)

## S4 method for signature 'tbl_df'
test_gene_rank(
    .data,
    .entrez,
    .arrange_desc,
    species,
    .sample = NULL,
    gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7"),
    gene_set = NULL
)

## S4 method for signature 'tidybulk'
test_gene_rank(
    .data,
    .entrez,
```r
.test_gene_rank

.data, .entrez, .arrange_desc, species, .sample = NULL, gene_sets = c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7"), gene_set = NULL

## S4 method for signature 'SummarizedExperiment'
.test_gene_rank(
  .data, .entrez, .arrange_desc, species, .sample = NULL, gene_sets = NULL, gene_set = NULL
)

## S4 method for signature 'RangedSummarizedExperiment'
.test_gene_rank(
  .data, .entrez, .arrange_desc, species, .sample = NULL, gene_sets = NULL, gene_set = NULL
)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.entrez The ENTREZ ID of the transcripts/genes

.arrange_desc A column name of the column to arrange in decreasing order

.species A character. For example, human or mouse. MSigDB uses the latin species names (e.g., "Mus musculus", "Homo sapiens")

.sample The name of the sample column

.gene_sets A character vector or a list. It can take one or more of the following built-in collections as a character vector: c("h", "c1", "c2", "c3", "c4", "c5", "c6", "c7", "kegg_disease", "kegg_metabolism", "kegg_signaling"), to be used with EGSEA buildIdx. c1 is human specific. Alternatively, a list of user-supplied gene sets can be provided, to be used with EGSEA buildCustomIdx. In that case, each gene set is a character vector of Entrez IDs and the names of the list are the gene set names.

.gene_set DEPRECATED. Use gene_sets instead.
**Details**

**[Maturing]**

This wrapper execute gene enrichment analyses of the dataset using a list of transcripts and GSEA. This wrapper uses clusterProfiler (DOI: doi.org/10.1089/omi.2011.0118) on the back-end.

Undelying method:

```r
# Get gene sets signatures
msigdb::msigdb(species = species)

# Filter specific gene_sets if specified. This was introduced to speed up examples executions when(!is.null(gene_sets) ~ filter(.., gs_cat ~ (.))) |

# Execute calculation
nest(data = -gs_cat) |> mutate(fit = map(data, ~ clusterProfiler::GSEA(my_entrez_rank, TERM2GENE = x) |> select(gs_name, entrez_gene), pvalueCutoff = 1))
```

**Value**

A consistent object (to the input)

- A `spec_tbl_df` object
- A `tbl_df` object
- A `tidybulk` object
- A `SummarizedExperiment` object
- A `RangedSummarizedExperiment` object

**Examples**

```r
print("Not run for build time.")

## Not run:

df_entrez = tidybulk::se_mini
df_entrez = mutate(df_entrez, do_test = .feature %in% c("TNFRSF4", "PLCH2", "PADI4", "PAX7"))
df_entrez = df_entrez |> test_differential_abundance(~ condition)

test_gene_rank(
  df_entrez,
  .sample = .sample,
  .entrez = entrez,
  species="Homo sapiens",
  gene_sets =c("C2"),
  .arrange_desc = logFC
)

## End(Not run)
```
test_stratification_cellularity

Test of stratification of biological replicates based on tissue composition, one cell-type at the time, using Kaplan-meier curves.

Description

test_stratification_cellularity() takes as input A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment)) and returns a consistent object (to the input) with additional columns for the statistics from the hypothesis test.

Usage

test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)

## S4 method for signature 'spec_tbl_df'
test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)

## S4 method for signature 'tbl_df'
test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)
## S4 method for signature 'tidybulk'

test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)

## S4 method for signature 'SummarizedExperiment'

test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)

## S4 method for signature 'RangedSummarizedExperiment'

test_stratification_cellularity(
  .data,
  .formula,
  .sample = NULL,
  .transcript = NULL,
  .abundance = NULL,
  method = "cibersort",
  reference = X_cibersort,
  ...
)

### Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.formula A formula representing the desired linear model. The formula can be of two forms: multivariable (recommended) or univariable Respectively: \"factor_of_interest \~ .\" or \". \~ factor_of_interest\". The dot represents cell-type proportions, and it is mandatory. If censored regression is desired (coxph) the formula should be of the form \"surv\n\n\nSurv(y, dead) \~ .\"
.sample The name of the sample column
.transcript The name of the transcript/gene column
.abundance The name of the transcript/gene abundance column

method A string character. Either "cibersort", "epic" or "llsr". The regression method will be chosen based on being multivariable: lm or cox-regression (both on logit-transformed proportions); or univariable: beta or cox-regression (on logit-transformed proportions). See .formula for multi- or univariable choice.

reference A data frame. The transcript/cell_type data frame of integer transcript abundance

Further parameters passed to the method deconvolve_cellularity

Details

'r lifecycle::badge("maturing")'

This routine applies a deconvolution method (e.g., Cibersort; DOI: 10.1038/nmeth.3337) and passes the proportions inferred into a generalised linear model (DOI:dx.doi.org/10.1007/s11749-010-0189-z) or a cox regression model (ISBN: 978-1-4757-3294-8)

Underlying method for the test: data |> deconvolve_cellularity( !!.sample, !!.transcript, !!.abundance, method=method, reference = reference, action="get", ... ) [...] |> mutate(.high_cellularity = .proportion > median(.proportion)) |> survival::survdiff(data = data, .my_formula)

Value

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

A consistent object (to the input) with additional columns for the statistics from the hypothesis test (e.g., log fold change, p-value and false discovery rate).

Examples

tidybulk::se_mini |> test_stratification_cellularity(
  survival::Surv(days, dead) ~ .,
  cores = 1
)

tidybulk

Creates an annotated ‘tidybulk’ tibble from a ‘tbl’ or ‘SummarizedExperiment’ object

Description

tidybulk() creates an annotated ‘tidybulk’ tibble from a ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

Usage

tidybulk(.data, .sample, .transcript, .abundance, .abundance_scaled = NULL)

Arguments

.data A ‘tbl’ (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

.sample The name of the sample column

.transcript The name of the transcript/gene column

.abundance The name of the transcript/gene abundance column

.abundance_scaled The name of the transcript/gene scaled abundance column

Details

‘r lifecycle::badge("maturing")’

This function creates a tidybulk object and is useful if you want to avoid to specify .sample, .transcript and .abundance arguments all the times. The tidybulk object have an attribute called internals where these three arguments are stored as metadata. They can be extracted as attr(<object>, "internals").
Value
A ‘tidybulk’ object
A ‘tidybulk’ object
A ‘tidybulk’ object
A ‘tidybulk’ object
A ‘tidybulk’ object
A ‘tidybulk’ object

Examples
```r
tidybulk(tidybulk::se_mini)
```

Description
tidybulk_SAM_BAM() creates a ‘tt’ object from a `tbl` (with at least three columns for sample, feature and transcript abundance) or ‘SummarizedExperiment’ (more convenient if abstracted to tibble with library(tidySummarizedExperiment))

Usage
tidybulk_SAM_BAM(file_names, genome = “hg38”, ...)

## S4 method for signature 'character,character'
tidybulk_SAM_BAM(file_names, genome = "hg38", ...)

Arguments
- `file_names`: A character vector
- `genome`: A character string specifying an in-built annotation used for read summarization. It has four possible values including "mm10", "mm9", "hg38" and "hg19"
- `...`: Further parameters passed to the function Rsubread::featureCounts

Details
```
'r lifecycle::badge("maturing")'
```
This function is based on FeatureCounts package (DOI: 10.1093/bioinformatics/btt656). This function creates a tidybulk object and is useful if you want to avoid to specify `.sample`, `.transcript` and `.abundance` arguments all the times. The tidybulk object have an attribute called internals where these three arguments are stored as metadata. They can be extracted as `attr(<object>, "internals")`. Underlying core function Rsubread::featureCounts(annot.inbuilt = genome,nthreads = n_cores, ...)
Value

A 'tidybulk' object
A 'tidybulk' object

---

tximeta_summarizeToGene_object

*Needed for tests tximeta_summarizeToGene_object, It is SummarizedExperiment from tximeta*

---

Description

Needed for tests tximeta_summarizeToGene_object, It is SummarizedExperiment from tximeta

Usage

```
tximeta_summarizeToGene_object
```

Format

An object of class RangedSummarizedExperiment with 10 rows and 1 columns.

---

unnest

unnest

---

Description

unnest

nest

Arguments

data A tbl. (See tidyr)
cols <[‘tidy-select’][tidyr_tidy_select]> Columns to unnest. If you ‘unnest()’ multiple columns, parallel entries must be of compatible sizes, i.e. they’re either equal or length 1 (following the standard tidyverse recycling rules).
names_sep If ‘NULL’, the default, the names will be left as is. In ‘nest()’, inner names will come from the former outer names; in ‘unnest()’, the new outer names will come from the inner names. If a string, the inner and outer names will be used together. In ‘nest()’, the names of the new outer columns will be formed by pasting together the outer and the inner column names, separated by ‘names_sep’. In ‘unnest()’, the new inner names will have the outer names (+ ‘names_sep’) automatically stripped. This makes ‘names_sep’ roughly symmetric between nesting and unnesting.
keep_empty  See tidyr::unnest
names_repair  See tidyr::unnest
ptype  See tidyr::unnest
.drop  See tidyr::unnest
.id  tidyr::unnest
.sep  tidyr::unnest
.preserve  See tidyr::unnest
.data  A tbl. (See tidyr)
...  Name-variable pairs of the form new_col = c(col1, col2, col3) (See tidyr)

Value

A tidySummarizedExperiment object or a tibble depending on input
A tt object

Examples

tidybulk::se_mini |> tidybulk() |> nest( data = -.feature) |> unnest(data)

tidybulk::se_mini %>% tidybulk() %>% nest( data = -.feature)

Description

Needed for vignette vignette_manuscript_signature_boxplot

Usage

vignette_manuscript_signature_boxplot

Format

An object of class tbl_df (inherits from tbl, data.frame) with 899 rows and 12 columns.
vignette_manuscript_signature_tsne

Description

Needed for vignette vignette_manuscript_signature_tsne

Usage

vignette_manuscript_signature_tsne

Format

An object of class spec_tbl_df (inherits from tbl_df, tbl, data.frame) with 283 rows and 10 columns.

vignette_manuscript_signature_tsne2

Description

Needed for vignette vignette_manuscript_signature_tsne2

Usage

vignette_manuscript_signature_tsne2

Format

An object of class tbl_df (inherits from tbl, data.frame) with 283 rows and 9 columns.

X_cibersort

Cibersort reference

Description

Cibersort reference

Usage

X_cibersort

Format

An object of class data.frame with 547 rows and 22 columns.
Pipe operator

**Description**

See `magrittr::%>%` for details.

**Usage**

```r
lhs %>% rhs
```

**Arguments**

- `lhs`: A value or the `magrittr` placeholder.
- `rhs`: A function call using the `magrittr` semantics.

**Value**

The result of calling `rhs(lhs)`. 
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