Package 'ISAnalytics'

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Title Analyze gene therapy vector insertion sites data identified from genomics next generation sequencing reads for clonal tracking studies

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Description In gene therapy, stem cells are modified using viral vectors to deliver the therapeutic transgene and replace functional properties since the genetic modification is stable and inherited in all cell progeny. The retrieval and mapping of the sequences flanking the virushost DNA junctions allows the identification of insertion sites (IS), essential for monitoring the evolution of genetically modified cells in vivo. A comprehensive toolkit for the analysis of IS is required to foster clonal trackign studies and supporting the assessment of safety and long term efficacy in vivo. This package is aimed at (1) supporting automation of IS workflow, (2) performing base and advance analysis for IS tracking (clonal abundance, clonal expansions and statistics for insertional mutagenesis, etc.), (3) providing basic biology insights of transduced stem cells in vivo.

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URL https://calabrialab.github.io/ISAnalytics,
 https://github.com//calabrialab/isanalytics

BugReports https://github.com/calabrialab/ISAnalytics/issues

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aggre	egate_metadata Performs aggregation on metadata contained in the association file.	

Description

[Stable] Groups metadata by the specified grouping keys and returns a summary of info for each group. For more details on how to use this function: vignette("aggregate_function_usage",package = "ISAnalytics")

Usage

```
aggregate_metadata(
  association_file,
  grouping_keys = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
  aggregating_functions = default_meta_agg(),
  import_stats = lifecycle::deprecated()
)
```

Arguments

association_file

The imported association file (via import_association_file)

grouping_keys A character vector of column names to form a grouping operation aggregating_functions

A data frame containing specifications of the functions to be applied to columns in the association file during aggregation. It defaults to default_meta_agg. The structure of this data frame should be maintained if the user wishes to change the defaults.

import_stats

[**Deprecated**] The import of VISPA2 stats has been moved to its dedicated function, see import_Vispa2_stats.

Value

An aggregated data frame

See Also

Other Aggregate functions: aggregate_values_by_key(), default_meta_agg()

Examples

```
data("association_file", package = "ISAnalytics")
aggreg_meta <- aggregate_metadata(
    association_file = association_file
)
head(aggreg_meta)</pre>
```

```
aggregate_values_by_key
```

Aggregates matrices values based on specified key.

Description

[**Stable**] Performs aggregation on values contained in the integration matrices based on the key and the specified lambda. For more details on how to use this function: vignette("aggregate_function_usage", package = "ISAnalytics")

Usage

```
aggregate_values_by_key(
    x,
    association_file,
    value_cols = "Value",
    key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    lambda = list(sum = ~sum(.x, na.rm = TRUE)),
    group = c(mandatory_IS_vars(), annotation_IS_vars()),
    join_af_by = "CompleteAmplificationID"
)
```

Arguments

x A single integration matrix or a list of imported integration matrices association_file

The imported association file

value_cols A character vector containing the names of the columns to apply the given lamb-

das. Must be numeric or integer columns.

key A string or a character vector with column names of the association file to take

as key

lambda A named list of functions or purrr-style lambdas. See details section.

group Other variables to include in the grouping besides key, can be set to NULL

join_af_by A character vector representing the joining key between the matrix and the meta-

data. Useful to re-aggregate already aggregated matrices.

Details

Setting the lambda parameter:

The lambda parameter should always contain a named list of either functions or purrr-style lambdas. It is also possible to specify the namespace of the function in both ways, for example:

```
lambda = list(sum = sum, desc = psych::describe)
```

Using purrr-style lambdas allows to specify arguments for the functions, keeping in mind that the first parameter should always be .x:

```
lambda = list(sum = ~sum(.x, na.rm = TRUE))
```

It is also possible to use custom user-defined functions, keeping in mind that the symbol will be evaluated in the calling environment, for example if the function is called in the global environment and lambda contains "foo" as a function, "foo" will be evaluated in the global environment.

```
foo <- function(x) {
   sum(x)
}
lambda = list(sum = ~sum(.x, na.rm = TRUE), foo = foo)
# Or with lambda notation
lambda = list(sum = ~sum(.x, na.rm = TRUE), foo = ~foo(.x))</pre>
```

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Constraints on aggregation functions:

Functions passed in the lambda parameters must respect a few constraints to properly work and it's the user responsibility to ensure this.

- Functions have to accept as input a numeric or integer vector
- Function should return a single value or a list/data frame: if a list or a data frame is returned as a result, all the columns will be added to the final data frame.

Value

A list of tibbles or a single tibble aggregated according to the specified arguments

See Also

```
Other Aggregate functions: aggregate_metadata(), default_meta_agg()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
head(aggreg)</pre>
```

annotation_issues

Check for genomic annotation problems in IS matrices.

Description

[Experimental] This helper function checks if each individual integration site, identified by the triplet (chr, integration locus, strand), has been annotated with two or more distinct gene symbols.

Usage

```
annotation_issues(matrix)
```

Arguments

matrix

Either a single matrix or a list of matrices, ideally obtained via import_parallel_Vispa2Matrices() or import_single_Vispa2Matrix()

Value

Either NULL if no issues were detected or 1 or more data frames with genomic coordinates of the IS and the number of distinct genes associated

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See Also

Other Import functions helpers: matching_options(), quantification_types()

Examples

```
data("integration_matrices", package = "ISAnalytics")
annotation_issues(integration_matrices)
```

annotation_IS_vars

Names of the annotation variables for an integration matrix.

Description

Contains the names of the columns that are present if the integration matrix is annotated.

Usage

```
annotation_IS_vars()
```

Value

A character vector

Examples

```
annotation_IS_vars()
```

association_file

Example of association file.

Description

This file is a simple example of association file. Use it as reference to properly fill out yours. To generate an empty association file to fill see the generate_blank_association_file() function.

Usage

```
data("association_file")
```

Format

An object of class data. table (inherits from data. frame) with 53 rows and 83 columns.

Details

The data was obtained manually by simulating real research data.

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See Also

```
generate_blank_association_file
```

```
association_file_columns
```

Names of the columns in the association file.

Description

All the names of the columns present in the association file.

Usage

```
association_file_columns()
```

Value

A character vector

Examples

```
association_file_columns()
```

as_sparse_matrix

Converts tidy integration matrices in the original sparse matrix form.

Description

[Stable] This function is particularly useful when a sparce matrix structure is needed by a specific function (mainly from other packages).

Usage

```
as_sparse_matrix(
    x,
    fragmentEstimate = "fragmentEstimate",
    seqCount = "seqCount",
    barcodeCount = "barcodeCount",
    cellCount = "cellCount",
    ShsCount = "ShsCount"
)
```

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Arguments

X	A single tidy integration matrix or a list of integration matrices. Supports also multi-quantification matrices obtained via comparison_matrix			
fragmentEstimate				
	For multi-quantification matrix support: the name of the fragment estimate values column			
seqCount	For multi-quantification matrix support: the name of the sequence count values column			
barcodeCount	For multi-quantification matrix support: the name of the barcode count values column			
cellCount	For multi-quantification matrix support: the name of the cell count values column			

Value

ShsCount

Depending on input, 2 possible outputs:

- A single sparce matrix (tibble) if input is a single quantification matrix
- A list of sparce matrices divided by quantification if input is a single multi-quantification matrix or a list of matrices

For multi-quantification matrix support: the name of the Shs Count values col-

See Also

```
Other Utility functions: generate_Vispa2_launch_AF(), generate_blank_association_file(), unzip_file_system()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
sparse <- as_sparse_matrix(integration_matrices)</pre>
```

```
available_outlier_tests
```

A character vector containing all the names of the currently supported outliers tests that can be called in the function outlier_filter.

Description

A character vector containing all the names of the currently supported outliers tests that can be called in the function outlier_filter.

Usage

```
available_outlier_tests()
```

Value

A character vector

See Also

```
Other Outlier tests: outliers_by_pool_fragments()
```

Examples

```
available_outlier_tests()
```

```
blood_lineages_default
```

Default blood lineages info

Description

A default table with info relative to different blood lineages associated with cell markers that can be supplied as a parameter to HSC_population_size_estimate

Usage

```
blood_lineages_default()
```

Value

A data frame

Examples

```
blood_lineages_default()
```

circos_genomic_density

Trace a circos plot of genomic densities.

Description

[Experimental] For this functionality the suggested package circlize is required. Please note that this function is a simple wrapper of basic circlize functions, for an in-depth explanation on how the functions work and additional arguments please refer to the official documentation Circular Visualization in R

Usage

```
circos_genomic_density(
  data,
  gene_labels = NULL,
  label_col = NULL,
  cytoband_specie = "hg19",
  track_colors = "navyblue",
  grDevice = c("png", "pdf", "svg", "jpeg", "bmp", "tiff", "default"),
  file_path = getwd(),
  ...
)
```

Arguments

data	Either a single integration matrix or a list of integration matrices. If a list is provided, a separate density track for each data frame is plotted.			
gene_labels	Either NULL or a data frame in bed format. See details.			
label_col	Numeric index of the column of gene_labels that contains the actual labels. Relevant only if gene_labels is not set to NULL.			
cytoband_specie				
	Specie for initializing the cytoband			
track_colors	Colors to give to density tracks. If more than one integration matrix is provided as data should be of the same length. Values are recycled if length of track_colors is smaller than the length of the input data.			
grDevice	The graphical device where the plot should be traced. default, if executing from RStudio is the viewer.			
file_path	If a device other than default is chosen, the path on disk where the file should be saved. Defaults to {current directory}/circos_plot.{device}.			
• • •	Additional named arguments to pass on to chosen device, circlize::circos.par(), circlize::circos.genomicDensity() and circlize::circos.genomicLabels()			

Details

Providing genomic labels:

If genomic labels should be plotted alongside genomic density tracks, the user should provide them as a simple data frame in standard bed format, namely chr, start, end plus a column containing the labels. NOTE: if the user decides to plot on the default device (viewer in RStudio), he must ensure there is enough space for all elements to be plotted, otherwise an error message is thrown.

Value

NULL

See Also

```
Other Plotting functions: CIS_volcano_plot(), HSC_population_plot(), integration_alluvial_plot(), sharing_heatmap(), sharing_venn(), top_abund_tableGrob()
```

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Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
by_subj <- aggreg %>%
    dplyr::group_by(.data$SubjectID) %>%
    dplyr::group_split()
circos_genomic_density(by_subj,
    track_colors = c("navyblue", "gold"),
    grDevice = "default", track.height = 0.1
)
```

CIS_grubbs

Grubbs test for Common Insertion Sites (CIS).

Description

[Stable] Statistical approach for the validation of common insertion sites significance based on the comparison of the integration frequency at the CIS gene with respect to other genes contained in the surrounding genomic regions. For more details please refer to this paper: https://ashpublications.org/blood/article/117/20/5332/21206/Lentiviral-vector-common-integration-sites-in

Usage

```
CIS_grubbs(
   x,
   genomic_annotation_file = "hg19",
   grubbs_flanking_gene_bp = 1e+05,
   threshold_alpha = 0.05,
   by = NULL
)
```

Arguments

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by

Either NULL or a character vector of column names. If not NULL, the function will perform calculations for each group and return a list of data frames with the results. E.g. for by = "SubjectID", CIS will be computed for each distinct SubjectID found in the table (of course, "SubjectID" column must be included in the input data frame).

Details

Genomic annotation file:

This file is a data base, or more simply a .tsv file to import, with genes annotation for the specific genome. The annotations for the human genome (hg19) and murine genome (mm9) are already included in this package: to use one of them just set the argument genomic_annotation_file to either "hg19" or "mm9". If for any reason the user is performing an analysis on another genome, this file needs to be changed respecting the USCS Genome Browser format, meaning the input file headers should include:

 $name 2\ chrom\ strand\ min_txStart\ max_txEnd\ minmax_TxLen\ average_TxLen\ name\ min_cdsStart\ max_cdsEnd\ minmax_CdsLen\ average_CdsLen$

Value

A data frame

See Also

```
Other Analysis functions: comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
cis <- CIS_grubbs(integration_matrices)
head(cis)</pre>
```

CIS_volcano_plot

Trace volcano plot for computed CIS data.

Description

[Stable] Traces a volcano plot for IS frequency and CIS results.

Usage

```
CIS_volcano_plot(
    x,
    onco_db_file = "proto_oncogenes",
    tumor_suppressors_db_file = "tumor_suppressors",
    species = "human",
```

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```
known_onco = known_clinical_oncogenes(),
suspicious_genes = clinical_relevant_suspicious_genes(),
significance_threshold = 0.05,
annotation_threshold_ontots = 0.1,
highlight_genes = NULL,
title_prefix = NULL,
return_df = FALSE
)
```

Arguments

x Either a simple integration matrix or a data frame resulting from the call to CIS_grubbs with add_standard_padjust = TRUE

onco_db_file Uniprot file for proto-oncogenes (see details). If different from default, should be supplied as a path to a file.

tumor_suppressors_db_file

Uniprot file for tumor-suppressor genes. If different from default, should be

supplied as a path to a file.

species One between "human", "mouse" and "all"

known_onco Data frame with known oncogenes. See details.

suspicious_genes

Data frame with clinical relevant suspicious genes. See details.

significance_threshold

The significance threshold

annotation_threshold_ontots

Value above which genes are annotated with colorful labels

highlight_genes

Either NULL or a character vector of genes to be highlighted in the plot even if

they're not above the threshold

title_prefix A string or character vector to be displayed in the title - usually the project name

and other characterizing info. If a vector is supplied, it is concatenated in a

single string via paste()

return_df Return the data frame used to generate the plot? This can be useful if the user

wants to manually modify the plot with ggplot2. If TRUE the function returns a

list containing both the plot and the data frame.

Details

Input data frame:

Users can supply as x either a simple integration matrix or a data frame resulting from the call to CIS_grubbs. In the first case an internal call to the function CIS_grubbs() is performed.

Oncogene and tumor suppressor genes files:

These files are included in the package for user convenience and are simply UniProt files with gene annotations for human and mouse. For more details on how this files were generated use the help ?tumor_suppressors, ?proto_oncogenes

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Known oncogenes:

The default values are included in this package and it can be accessed by doing:

head(known_clinical_oncogenes())

```
## # A tibble: 5 × 2
    GeneName KnownClonalExpansion
##
##
    <chr>
              <lg1>
## 1 MECOM
              TRUE
## 2 CCND2
              TRUE
## 3 TAL1
              TRUE
## 4 LMO2
              TRUE
## 5 HMGA2
              TRUE
```

If the user wants to change this parameter the input data frame must preserve the column structure. The same goes for the suspicious_genes parameter (DOIReference column is optional):

head(clinical_relevant_suspicious_genes())

```
## # A tibble: 6 × 3
##
     GeneName ClinicalRelevance DOIReference
##
     <chr>
              <1g1>
## 1 DNMT3A
              TRUE
                                https://doi.org/10.1182/blood-2018-01-829937
## 2 TET2
              TRUE
                                https://doi.org/10.1182/blood-2018-01-829937
## 3 ASXL1
              TRUE
                                https://doi.org/10.1182/blood-2018-01-829937
## 4 JAK2
              TRUF
                                https://doi.org/10.1182/blood-2018-01-829937
## 5 CBL
              TRUE
                                https://doi.org/10.1182/blood-2018-01-829937
## 6 TP53
              TRUE
                                https://doi.org/10.1182/blood-2018-01-829937
```

Value

A plot or a list containing a plot and a data frame

See Also

```
Other Plotting functions: HSC_population_plot(), circos_genomic_density(), integration_alluvial_plot(), sharing_heatmap(), sharing_venn(), top_abund_tableGrob()
```

Examples

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```
clinical_relevant_suspicious_genes
```

Clinical relevant suspicious genes (for mouse and human).

Description

Clinical relevant suspicious genes (for mouse and human).

Usage

```
clinical_relevant_suspicious_genes()
```

Value

A data frame

See Also

Other Plotting function helpers: known_clinical_oncogenes()

Examples

```
clinical_relevant_suspicious_genes()
```

comparison_matrix

obtain a single integration matrix from individual quantification matrices.

Description

[Stable] Takes a list of integration matrices referring to different quantification types and merges them in a single data frame that has multiple value columns, each renamed according to their quantification type of reference.

Usage

```
comparison_matrix(
    x,
    fragmentEstimate = "fragmentEstimate",
    seqCount = "seqCount",
    barcodeCount = "barcodeCount",
    cellCount = "cellCount",
    ShsCount = "ShsCount"
)
```

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Arguments

x A named list of integration matrices, ideally obtained via import_parallel_Vispa2Matrices_interactive or import_parallel_Vispa2Matrices_auto. Names must be quantification types.

fragmentEstimate

The name of the output column for fragment estimate values

seqCount The name of the output column for sequence count values
barcodeCount The name of the output column for barcode count values
cellCount The name of the output column for cell count values
ShsCount The name of the output column for Shs count values

Value

A tibble

See Also

```
quantification_types
```

```
Other Analysis functions: CIS_grubbs(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

compute_abundance

Computes the abundance for every integration event in the input data frame.

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Description

[Stable] Abundance is obtained for every integration event by calculating the ratio between the single value and the total value for the given group.

Usage

```
compute_abundance(
    x,
    columns = c("fragmentEstimate_sum"),
    percentage = TRUE,
    key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    keep_totals = FALSE
)
```

Arguments

x An integration matrix - aka a data frame that includes the mandatory_IS_vars()

as columns. The matrix can either be aggregated (via aggregate_values_by_key())

or not.

columns A character vector of column names to process, must be numeric or integer

columns

percentage Add abundance as percentage?

key The key to group by when calculating totals

keep_totals A value between TRUE, FALSE or df. If TRUE, the intermediate totals for each

group will be kept in the output data frame as a dedicated column with a trailing "_tot". If FALSE, totals won't be included in the output data frame. If df, the totals are returned to the user as a separate data frame, together with the

abundance data frame.

Details

Abundance will be computed upon the user selected columns in the columns parameter. For each column a corresponding relative abundance column (and optionally a percentage abundance column) will be produced.

Value

Either a single data frame with computed abundance values or a list of 2 data frames (abundance_df, quant_totals)

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
abund <- compute_abundance(
    x = integration_matrices,
    columns = "fragmentEstimate",
    key = "CompleteAmplificationID"
)
head(abund)</pre>
```

compute_near_integrations

Scans input matrix to find and merge near integration sites.

Description

[Stable] This function scans the input integration matrix to detect eventual integration sites that are too "near" to each other and merges them into single integration sites adjusting their values if needed.

Usage

```
compute_near_integrations(
    x,
    threshold = 4,
    keep_criteria = "max_value",
    strand_specific = TRUE,
    value_columns = c("seqCount", "fragmentEstimate"),
    max_value_column = "seqCount",
    map_as_file = TRUE,
    file_path = default_report_path()
)
```

Arguments

An integration matrix

threshold

A single integer that represents an absolute number of bases for which two integrations are considered distinct. If the threshold is set to 3 it means, provided fields chr and strand are the same, integrations sites which have at least 3 bases in between them are considered distinct (e.g. (1, 14576, +) and (1, 14580, +) are considered distinct)

keep_criteria

While scanning, which integration should be kept? The 2 possible choices for this parameter are:

- "max_value": keep the integration site which has the highest value (and collapse other values on that integration).
- "keep_first": keeps the first integration

strand_specific

Should strand be considered? If yes, for example these two integration sites (chr = "1", strand = "+", integration_locus = 14568) and (chr = "1", strand = "-", integration_locus = 14568) are considered different and not grouped together.

value_columns Character vector, contains the names of the numeric experimental columns max_value_column

The column that has to be considered for searching the maximum value

map_as_file Produce recalibration map as a .tsv file?

file_path String representing the path were the file will be saved. Can be either a folder or

a file. Relevant only if map_as_file is TRUE.

Details

The whole matrix is scanned with a sliding window mechanism: for each row in the integration matrix an interval is calculated based on the threshold value, then a "look ahead" operation is performed to detect subsequent rows which integration locuses fall in the interval. If CompleteAmplificationIDs of the near integrations are different only the locus value (and optionally GeneName and GeneStrand if the matrix is annotated) is modified, otherwise rows with the same id are aggregated and values are summed. The function will also produce a re-calibration map: this data frame contains the reference of pre-recalibration values for chr, strand and integration_locus and the value to which that integration was changed to.

Value

An integration matrix with same or less number of rows

Note

We do recommend to use this function in combination with comparison_matrix to automatically perform re-calibration on all quantification matrices.

Examples

```
data("integration_matrices", package = "ISAnalytics")
rec <- compute_near_integrations(
    x = integration_matrices, map_as_file = FALSE
)
head(rec)</pre>
```

cumulative_count_union

Integrations cumulative count in time by sample

Description

[Experimental] This function computes the cumulative number of integrations observed in each sample at different time points by assuming that if an integration is observed at time point "t" then it is also observed in time point "t+1".

Usage

```
cumulative_count_union(
    x,
    association_file = NULL,
    timepoint_column = "TimePoint",
    key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    include_tp_zero = FALSE,
    zero = "0000",
    aggregate = FALSE,
    ...
)
```

Arguments

Details

Input data frame:

The user can provide as input for the x parameter both a simple integration matrix AND setting the aggregate parameter to TRUE, or provide an already aggregated matrix via aggregate_values_by_key. If the user supplies a matrix to be aggregated the association_file parameter must not be NULL: aggregation will be done by an internal call to the aggregation function. If the user supplies an already aggregated matrix, the key parameter is the key used for aggregation - NOTE: for this operation is mandatory that the time point column is included in the key.

Assumptions on time point format:

By using the functions provided by this package, when imported, an association file will be correctly formatted for future usage. In the formatting process there is also a padding operation performed on time points: this means the functions expects the time point column to be of type character and to be correctly padded with 0s. If the chosen column for time point is detected as numeric the function will attempt the conversion to character and automatic padding. If you choose to import the association file not using the import_association_file function, be sure to check the format of the chosen column to avoid undesired results.

Value

A data frame

22 cumulative_is

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
cumulative_count <- cumulative_count_union(aggreg)
cumulative_count</pre>
```

cumulative_is

Expands integration matrix with the cumulative is union over time.

Description

[Experimental] Given an input integration matrix that can be grouped over time, this function adds integrations in groups assuming that if an integration is observed at time point "t" then it is also observed in time point "t+1".

Usage

```
cumulative_is(
    x,
    key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    timepoint_col = "TimePoint",
    include_tp_zero = FALSE,
    keep_og_is = TRUE,
    expand = FALSE
)
```

Arguments

```
An integration matrix, ideally aggregated via aggregate_values_by_key()

key The aggregation key used

timepoint_col The name of the time point column

include_tp_zero

Should time point 0 be included?

keep_og_is Keep original set of integrations as a separate column?

expand If FALSE, for each group, the set of integration sites is returned in a separate column as a nested table, otherwise the resulting column is unnested.
```

date_columns_coll 23

Value

A data frame

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
cumulated_is <- cumulative_is(aggreg)
cumulated_is</pre>
```

date_columns_coll

Possible choices for date_col parameter.

Description

Possible choices for date_col parameter.

Usage

```
date_columns_coll()
```

Value

A character vector of column names

See Also

```
remove_collisions
```

Examples

```
dates <- date_columns_coll()</pre>
```

24 date_formats

Description

All options correspond to lubridate functions:

• ymd: year, month, date

• ydm: year, day, month

• mdy: month, day, year

• myd: month, year, day

• dmy: day, month, year

• dym: day, year, month

• yq: year quantile

Usage

```
date_formats()
```

Details

NOTE: use the same date format across the association file.

Value

A character vector

See Also

```
import_association_file, import_parallel_Vispa2Matrices_auto
```

Examples

```
date_formats()
```

default_iss_file_prefixes

```
default_iss_file_prefixes
```

Default regex prefixes for Vispa2 stats files.

Description

Note that each element is a regular expression.

Usage

```
default_iss_file_prefixes()
```

Value

A character vector of regexes

Examples

```
default_iss_file_prefixes()
```

default_meta_agg

Default metadata aggregation function table

Description

A default columns-function specifications for aggregate_metadata

Usage

```
default_meta_agg()
```

Details

This data frame contains four columns:

- Column: holds the name of the column in the association file that should be processed
- Function: contains either the name of a function (e.g. mean) or a purrr-style lambda (e.g. ~ mean(.x,na.rm = TRUE)). This function will be applied to the corresponding column specified in Column
- Args: optional additional arguments to pass to the corresponding function. This is relevant ONLY if the corresponding Function is a simple function and not a purrr-style lambda.
- Output_colname: a glue specification that will be used to determine a unique output column name. See glue for more details.

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Value

A data frame

See Also

Other Aggregate functions: aggregate_metadata(), aggregate_values_by_key()

Examples

```
default_meta_agg()
```

default_report_path

Default folder for saving ISAnalytics reports. Supplied as default argument for several functions.

Description

Default folder for saving ISAnalytics reports. Supplied as default argument for several functions.

Usage

```
default_report_path()
```

Value

A path

Examples

```
default_report_path()
```

 $default_stats$

A set of pre-defined functions for sample_statistics.

Description

A set of pre-defined functions for sample_statistics.

Usage

```
default_stats()
```

Value

A named list of functions/purrr-style lambdas

Examples

```
default_stats()
```

```
generate_blank_association_file
```

Creates a blank association file.

Description

This function is useful if you want a blank association file to start using both Vispa2 and this package or simply if you want a correct framework to fix a malformed association file you have already.

Usage

```
generate_blank_association_file(path)
```

Arguments

path

The path on disk where the file should be written

Value

returns NULL

See Also

```
Other Utility functions: as_sparse_matrix(), generate_Vispa2_launch_AF(), unzip_file_system()
```

Examples

```
temp <- tempfile()
generate_blank_association_file(temp)</pre>
```

```
generate_Vispa2_launch_AF
```

Creates a reduced association file for Vispa2 run, given project and pool

Description

The function selects the appropriate columns and prepares a file for the launch of Vispa2 pipeline for each project/pool pair specified.

Usage

```
generate_Vispa2_launch_AF(association_file, project, pool, path)
```

Arguments

association_file

The imported association file (via import_association_file)

project A vector of characters containing project names

pool A vector of characters containing pool names. **NOTE: the names should refer**

to the values contained in the PoolID column of the association file and NOT

the concatenatePoolIDSeqRun column!

path A single string representing the path to the folder where files should be written.

If the folder doesn't exist it will be created.

Details

Note: the function is vectorized, meaning you can specify more than one project and more than one pool as vectors of characters, but you must ensure that:

- Both project and pool vectors have the same length
- You correctly type names in corresponding positions, for example c("CLOEXP", "PROJECT1100",
 "PROJECT1100") c("POOL6", "ABX-LR-PL5-POOL14-1", "ABX-LR-PL6-POOL15-1").
 If you type a pool in the position of a corresponding project that doesn't match no file will be produced since that pool doesn't exist in the corresponding project.

Value

returns NULL

See Also

```
Other Utility functions: as_sparse_matrix(), generate_blank_association_file(), unzip_file_system()
```

Examples

```
temp <- tempdir()
data("association_file", package = "ISAnalytics")
generate_Vispa2_launch_AF(association_file, "PJ01", "POOL01", temp)</pre>
```

HSC_population_plot

Plot of the estimated HSC population size for each patient.

Description

Plot of the estimated HSC population size for each patient.

Usage

```
HSC_population_plot(
  estimates,
  project_name,
  timepoints = "Consecutive",
  models = "Mth Chao (LB)"
)
```

Arguments

roject_name The estimates data frame, obtained via HSC_population_size_estimate

The project name, will be included in the plot title

Which time points to plot? One between "All", "Stable" and "Consecutive"

Name of the models to plot (as they appear in the column of the estimates)

Value

A plot

See Also

```
Other Plotting functions: CIS_volcano_plot(), circos_genomic_density(), integration_alluvial_plot(), sharing_heatmap(), sharing_venn(), top_abund_tableGrob()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
aggreg_meta <- aggregate_metadata(
    association_file = association_file
)
estimate <- HSC_population_size_estimate(
    x = aggreg,
    metadata = aggreg_meta,
    stable_timepoints = c(90, 180, 360),
    cell_type = "Other"
)
p <- HSC_population_plot(estimate, "PJ01")</pre>
```

```
HSC_population_size_estimate
```

Hematopoietic stem cells population size estimate.

Description

[Experimental] Hematopoietic stem cells population size estimate with capture-recapture models.

Usage

```
HSC_population_size_estimate(
    x,
    metadata,
    stable_timepoints = NULL,
    aggregation_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    blood_lineages = blood_lineages_default(),
    timepoint_column = "TimePoint",
    seqCount_column = "seqCount_sum",
    fragmentEstimate_column = "fragmentEstimate_sum",
    seqCount_threshold = 3,
    fragmentEstimate_threshold = 3,
    nIS_threshold = 5,
    cell_type = "MYELOID",
    tissue_type = "PB"
)
```

Arguments

x An aggregated integration matrix. See details.

metadata An aggregated association file. See details.

 $stable_timepoints$

A numeric vector or NULL if there are no stable time points.

aggregation_key

A character vector indicating the key used for aggregating x and metadata. Note that x and metadata should always be aggregated with the same key.

blood_lineages A data frame containing information on the blood lineages. Users can supply their own, provided the columns CellMarker and CellType are present.

timepoint_column

What is the name of the time point column to use? Note that this column must be present in the key.

seqCount_column

What is the name of the column in x containing the values of sequence count quantification?

fragmentEstimate_column

What is the name of the column in x containing the values of fragment estimate quantification? If fragment estimate is not present in the matrix, param should be set to NULL.

seqCount_threshold

A single numeric value. After re-aggregating x, rows with a value greater or equal will be kept, the others will be discarded.

fragmentEstimate_threshold

A single numeric value. Threshold value for fragment estimate, see details.

nIS_threshold A single numeric value. If a group (row) in the metadata data frame has a count

of distinct integration sites strictly greater than this number it will be kept, oth-

erwise discarded.

cell_type The cell types to include in the models. Note that the matching is case-insensitive.

tissue_type The tissue types to include in the models. Note that the matching is case-

insensitive.

Value

A data frame with the results of the estimates

Input formats

Both x and metadata should be supplied to the function in aggregated format (ideally through the use of aggregate_metadata and aggregate_values_by_key). Note that the aggregation_key, aka the vector of column names used for aggregation, must contain at least the columns SubjectID, CellMarker, Tissue and a time point column (the user can specify the name of the column in the argument timepoint_column).

On time points

If stable_timepoints is a vector with length > 1, the function will look for the first available stable time point and slice the data from that time point onward. If NULL is supplied instead, it means there are no stable time points available. Note that 0 time points are ALWAYS discarded. Also, to be included in the analysis, a group must have at least 2 distinct non-zero time points.

Setting a threshold for fragment estimate

If fragment estimate is present in the input matrix, the filtering logic changes slightly: rows in the original matrix are kept if the sequence count value is greater or equal than the seqCount_threshold AND the fragment estimate value is greater or equal to the fragmentEstimate_threshold IF PRESENT (non-zero value). This means that for rows that miss fragment estimate, the filtering logic will be applied only on sequence count. If the user wishes not to use the combined filtering with fragment estimate, simply set fragmentEstimate_threshold = 0.

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(</pre>
```

```
x = integration_matrices,
   association_file = association_file,
   value_cols = c("seqCount", "fragmentEstimate")
)
aggreg_meta <- aggregate_metadata(association_file = association_file)
estimate <- HSC_population_size_estimate(
   x = aggreg,
   metadata = aggreg_meta,
   fragmentEstimate_column = NULL,
   stable_timepoints = c(90, 180, 360),
   cell_type = "Other"
)</pre>
```

import_association_file

Import the association file from disk

Description

[Stable] Imports the association file and immediately performs a check on the file system starting from the root to assess the alignment between the two.

Usage

```
import_association_file(
  path,
  root = NULL,
  tp_padding = 4,
  dates_format = "ymd",
  separator = "\t",
  filter_for = NULL,
  import_iss = FALSE,
  convert_tp = TRUE,
  report_path = default_report_path(),
  ...
)
```

Arguments

path The path on disk to the association file.

root The path on disk of the root folder of Vispa2 output or NULL. See details.

tp_padding Timepoint padding, indicates the number of digits of the "TimePoint" column

once imported. Fills the content with 0s up to the length specified (ex: 1 be-

comes 0001 with a tp_padding of 4)

dates_format A single string indicating how dates should be parsed. Must be a value in:

date_formats()

separator The column separator used in the file

filter_for	A named list where names represent column names that must be filtered. For example: $list(ProjectID = c("PROJECT1", "PROJECT2))$ will filter the association file so that it contains only those rows for which the value of the column "ProjectID" is one of the specified values. If multiple columns are present in the list all filtering conditions are applied as a logical AND.
import_iss	Import Vispa2 stats and merge them with the association file?
convert_tp	Should be time points be converted into months and years?
report_path	The path where the report file should be saved. Can be a folder, a file or NULL if no report should be produced. Defaults to {user_home}/ISAnalytics_reports.
	Additional arguments to pass to import_Vispa2_stats

Details

If the root argument is set to NULL no file system alignment is performed. This allows to import the basic file but it won't be possible to perfom automated matrix and stats import. For more details see the "How to use import functions" vignette: vignette("import_functions_howto",package = "ISAnalytics")

Value

The data frame holding metadata

See Also

```
date_formats
```

Other Import functions: import_Vispa2_stats(), import_parallel_Vispa2Matrices(), import_single_Vispa2Matri

Examples

```
fs_path <- system.file("extdata", "fs.zip", package = "ISAnalytics")
fs <- unzip_file_system(fs_path, "fs")
af_path <- system.file("extdata", "asso.file.tsv.gz", package = "ISAnalytics")
af <- import_association_file(af_path, root = fs, report_path = NULL)
head(af)</pre>
```

```
import_parallel_Vispa2Matrices
```

Import integration matrices from paths in the association file.

Description

[Stable] The function offers a convenient way of importing multiple integration matrices in an automated or semi-automated way. For more details see the "How to use import functions" vignette: vignette("import_functions_howto",package = "ISAnalytics")

Usage

```
import_parallel_Vispa2Matrices(
   association_file,
   quantification_type,
   matrix_type = "annotated",
   workers = 2,
   multi_quant_matrix = TRUE,
   report_path = default_report_path(),
   patterns = NULL,
   matching_opt = matching_options(),
   mode = c("AUTO", "INTERACTIVE"),
   ...
)
```

Arguments

association_file

Data frame imported via import_association_file (with file system alignment) or a string containing the path to the association file on disk.

quantification_type

A vector of requested quantification_types. Possible choices are quantifica-

tion_types

matrix_type A single string representing the type of matrices to be imported. Can only be

one in "annotated" or "not_annotated".

workers A single integer representing the number of parallel workers to use for the import

multi_quant_matrix

If set to TRUE will produce a multi-quantification matrix through compari-

son matrix instead of a list.

report_path The path where the report file should be saved. Can be a folder, a file or NULL if

no report should be produced. Defaults to {user_home}/ISAnalytics_reports.

patterns Relevant only if argument mode is set to AUTO. A character vector of additional

patterns to match on file names. Please note that patterns must be regular ex-

pressions. Can be NULL if no patterns need to be matched.

matching_opt Relevant only if argument mode is set to AUTO. A single value between match-

ing_options

mode A single value between AUTO and INTERACTIVE. If INTERACTIVE, the func-

tion will ask for input from the user on console, otherwise the process is fully

automated (with limitations, see vignette).

.. <dynamic-dots> Additional named arguments to pass to import_association_file

and comparison_matrix

Value

Either a multi-quantification matrix or a list of integration matrices

See Also

Other Import functions: import_Vispa2_stats(), import_association_file(), import_single_Vispa2Matrix()

Examples

import_single_Vispa2Matrix

Import a single integration matrix from file

Description

[Stable] This function allows to read and import an integration matrix produced as the output of Vispa2 pipeline and converts it to a tidy format.

Usage

```
import_single_Vispa2Matrix(
  path,
  to_exclude = NULL,
  keep_excluded = FALSE,
  separator = "\t"
)
```

Arguments

path The path to the file on disk

to_exclude Either NULL or a character vector of column names that should be ignored when

importing

keep_excluded Keep the columns in to_exclude as additional id columns?

separator The column delimiter used, defaults to \t

Details

For more details see the "How to use import functions" vignette: vignette("import_functions_howto", package = "ISAnalytics")

Value

A data.table object in tidy format

See Also

Other Import functions: import_Vispa2_stats(), import_association_file(), import_parallel_Vispa2Matrices()

Examples

```
fs_path <- system.file("extdata", "fs.zip", package = "ISAnalytics")
fs <- unzip_file_system(fs_path, "fs")
matrix_path <- fs::path(
    fs,
        "PJ01",
        "quantification",
        "POOL01-1",
        "PJ01_POOL01-1_seqCount_matrix.no0.annotated.tsv.gz"
)
matrix <- import_single_Vispa2Matrix(matrix_path)
head(matrix)</pre>
```

Description

[Stable] Imports all the Vispa2 stats files for each pool provided the association file has been aligned with the file system (see import_association_file).

Usage

```
import_Vispa2_stats(
   association_file,
   file_prefixes = default_iss_file_prefixes(),
   join_with_af = TRUE,
   pool_col = "concatenatePoolIDSeqRun",
   report_path = default_report_path()
)
```

Arguments

association_file

The file system aligned association file (contains columns with absolute paths to

the 'iss' folder)

file_prefixes A character vector with known file prefixes to match on file names. NOTE: the

elements represent regular expressions. For defaults see default_iss_file_prefixes.

join_with_af Logical, if TRUE the imported stats files will be merged with the association

file, if false a single data frame holding only the stats will be returned.

pool_col A single string. What is the name of the pool column used in the Vispa2 run?

This will be used as a key to perform a join operation with the stats files POOL

column.

report_path The path where the report file should be saved. Can be a folder, a file or NULL if

no report should be produced. Defaults to {user_home}/ISAnalytics_reports.

Value

A data frame

See Also

```
Other Import functions: import_association_file(), import_parallel_Vispa2Matrices(), import_single_Vispa2Matrix()
```

Examples

integration_alluvial_plot

Alluvial plots for IS distribution in time.

Description

[Experimental] Alluvial plots allow the visualization of integration sites distribution in different points in time in the same group. This functionality requires the suggested package ggalluvial.

Usage

```
integration_alluvial_plot(
    x,
    group = c("SubjectID", "CellMarker", "Tissue"),
    plot_x = "TimePoint",
    plot_y = "fragmentEstimate_sum_PercAbundance",
    alluvia = mandatory_IS_vars(),
    alluvia_plot_y_threshold = 1,
    top_abundant_tbl = TRUE,
    ...
)
```

Arguments

X	A data frame. See details.
group	Character vector containing the column names that identify unique groups.
plot_x	Column name to plot on the x axis
plot_y	Column name to plot on the y axis
alluvia	Character vector of column names that uniquely identify alluvia
alluvia_plot_y_threshold	
	Numeric value. Everything below this threshold on y will be plotted in grey and aggregated. See details.
top_abundant_tbl	
	Logical. Produce the summary top abundant tables via top_abund_tableGrob?
	Additional arguments to pass on to top_abund_tableGrob

Details

Input data frame:

The input data frame must contain all the columns specified in the arguments group, plot_x, plot_y and alluvia. The standard input for this function is the data frame obtained via the compute abundance function.

Plotting threshold on y:

The plotting threshold on the quantification on the y axis has the function to highlight only relevant information on the plot and reduce computation time. The default value is 1, that acts on the default column plotted on the y axis which holds a percentage value. This translates in natural language roughly as "highlight with colors only those integrations (alluvia) that at least in 1 point in time have an abundance value >= 1 %". The remaining integrations will be plotted as transparent in the strata.

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Value

For each group a list with the associated plot and optionally the summary tableGrob

See Also

```
Other Plotting functions: CIS_volcano_plot(), HSC_population_plot(), circos_genomic_density(), sharing_heatmap(), sharing_venn(), top_abund_tableGrob()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(</pre>
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
abund <- compute_abundance(x = aggreg)</pre>
alluvial_plots <- integration_alluvial_plot(abund,</pre>
    alluvia_plot_y_threshold = 0.5
ex_plot <- alluvial_plots[[1]]$plot +
    ggplot2::labs(
        title = "IS distribution over time",
        subtitle = "Patient 1, MNC BM",
        y = \text{"Abundance (\%)"},
        x = "Time point (days after GT)"
print(ex_plot)
```

 $integration_matrices \quad \textit{Example of imported multi-quantification integration matrices}.$

Description

The data was obtained manually by simulating real research data.

Usage

```
data("integration_matrices")
```

Format

Data frame with 1689 rows and 8 columns

chr The chromosome number (as character)

integration_locus Number of the base at which the viral insertion occurred
strand Strand of the integration

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GeneName Symbol of the closest gene

GeneStrand Strand of the closest gene

CompleteAmplificationID Unique sample identifier

seqCount Value of the sequence count quantification

fragmentEstimate Value of the fragment estimate quantification

iss_source

Find the source of IS by evaluating sharing.

Description

[Experimental] The function computes the sharing between a reference group of interest for each time point and a selection of groups of interest. In this way it is possible to observe the percentage of shared integration sites between reference and each group and identify in which time point a certain IS was observed for the first time.

Usage

```
iss_source(
  reference,
  selection,
  ref_group_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
  selection_group_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
  timepoint_column = "TimePoint",
  by_subject = TRUE,
  subject_column = "SubjectID"
)
```

Arguments

reference A data frame containing one or more groups of reference. Groups are identified

by ref_group_key

selection A data frame containing one or more groups of interest to compare. Groups are

identified by selection_group_key

ref_group_key Character vector of column names that identify a unique group in the reference

data frame

selection_group_key

Character vector of column names that identify a unique group in the selection

data frame

timepoint_column

Name of the column holding time point info?

by_subject Should calculations be performed for each subject separately?

subject_column Name of the column holding subjects information. Relevant only if by_subject

= TRUE

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Value

A list of data frames or a data frame

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(</pre>
   x = integration_matrices,
   association_file = association_file,
   value_cols = c("seqCount", "fragmentEstimate")
)
df1 <- aggreg %>%
   dplyr::filter(.data$Tissue == "BM")
df2 <- aggreg %>%
    dplyr::filter(.data$Tissue == "PB")
source <- iss_source(df1, df2)</pre>
source
ggplot2::ggplot(source$PT001, ggplot2::aes(
    x = as.factor(g2_TimePoint),
   y = sharing_perc, fill = g1
)) +
    ggplot2::geom_col() +
    ggplot2::labs(
        x = "Time point", y = "Shared IS % with MNC BM",
        title = "Source of is MNC BM vs MNC PB"
   )
```

is_sharing

Sharing of integration sites between given groups.

Description

[Experimental] Computes the amount of integration sites shared between the groups identified in the input data.

Usage

```
is_sharing(
...,
  group_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
  group_keys = NULL,
  n_comp = 2,
```

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```
is_count = TRUE,
relative_is_sharing = TRUE,
minimal = TRUE,
include_self_comp = FALSE,
keep_genomic_coord = FALSE,
table_for_venn = FALSE
```

Arguments

... One or more integration matrices
group_key Character vector of column names which id

group_key Character vector of column names which identify a single group. An associated

group id will be derived by concatenating the values of these fields, separated

by "_"

group_keys A list of keys for asymmetric grouping. If not NULL the argument group_key

is ignored

n_comp Number of comparisons to compute. This argument is relevant only if provided

a single data frame and a single key.

is_count Logical, if TRUE returns also the count of IS for each group and the count for the

union set

relative_is_sharing

Logical, if TRUE also returns the relative sharing.

minimal Compute only combinations instead of all possible permutations? If TRUE saves

time and excludes redundant comparisons.

include_self_comp

Include comparisons with the same group?

keep_genomic_coord

If TRUE keeps the genomic coordinates of the shared integration sites in a dedi-

cated column (as a nested table)

table_for_venn Add column with truth tables for venn plots?

Details

An integration site is always identified by the triple (chr, integration_locus, strand), thus these columns must be present in the input(s).

The function accepts multiple inputs for different scenarios, please refer to the vignette vignette("sharing_analyses",pac = "ISAnalytics") for a more in-depth explanation.

Output:

The function outputs a single data frame containing all requested comparisons and optionally individual group counts, genomic coordinates of the shared integration sites and truth tables for plotting venn diagrams.

Plotting sharing:

The sharing data obtained can be easily plotted in a heatmap via the function sharing_heatmap or via the function sharing_venn

Value

A data frame

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
sharing <- is_sharing(aggreg)
sharing</pre>
```

known_clinical_oncogenes

Known clinical oncogenes (for mouse and human).

Description

Known clinical oncogenes (for mouse and human).

Usage

```
known_clinical_oncogenes()
```

Value

A data frame

See Also

Other Plotting function helpers: clinical_relevant_suspicious_genes()

```
known_clinical_oncogenes()
```

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mandatory_IS_vars

Names of mandatory variables for an integration matrix.

Description

Contains the names of the columns that need to be present in order for a tibble to be considered an integration matrix.

Usage

```
mandatory_IS_vars()
```

Value

A character vector

Examples

mandatory_IS_vars()

matching_options

Possible choices for the matching_opt parameter.

Description

These are all the possible values for the matching_opt parameter in import_parallel_vispa2Matrices_auto.

Usage

```
matching_options()
```

Details

The values "ANY", "ALL" and "OPTIONAL", represent how the patterns should be matched, more specifically

- ANY = look only for files that match AT LEAST one of the patterns specified
- ALL = look only for files that match ALL of the patterns specified
- OPTIONAL = look preferentially for files that match, in order, all patterns or any pattern and if no match is found return what is found (keep in mind that duplicates are discarded in automatic mode)

Value

A vector of characters for matching_opt

See Also

```
import_parallel_Vispa2Matrices_auto
Other Import functions helpers: annotation_issues(), quantification_types()
```

Examples

```
opts <- matching_options()</pre>
```

```
outliers_by_pool_fragments
```

Identify and flag outliers based on pool fragments.

Description

[Experimental] Identify and flag outliers

Usage

```
outliers_by_pool_fragments(
  metadata,
  key = "BARCODE_MUX",
  outlier_p_value_threshold = 0.01,
  normality_test = FALSE,
  normality_p_value_threshold = 0.05,
  transform_log2 = TRUE,
  per_pool_test = TRUE,
  pool_col = "PoolID",
  min_samples_per_pool = 5,
  flag_logic = "AND",
  keep_calc_cols = TRUE,
  report_path = default_report_path()
)
```

Arguments

```
metadata The metadata data frame

key A character vector of numeric column names

outlier_p_value_threshold

The p value threshold for a read to be considered an outlier

normality_test Perform normality test? Normality is assessed for each column
```

normality_test Perform normality test? Normality is assessed for each column in the key using Shapiro-Wilk test and if the values do not follow a normal distribution, other calculations are skipped

normality_p_value_threshold

Normality threshold

transform_log2 Perform a log2 trasformation on values prior the actual calculations?

per_pool_test Perform the test for each pool?

pool_col A character vector of the names of the columns that uniquely identify a pool min_samples_per_pool

The minimum number of samples that a pool needs to contain in order to be processed - relevant only if per_pool_test = TRUE

flag_logic A character vector of logic operators to obtain a global flag formula - only relevant if the key is longer than one. All operators must be chosen between: AND, OR, XOR, NAND, NOR, XNOR

keep_calc_cols Keep the calculation columns in the output data frame?

report_path The path where the report file should be saved. Can be a folder, a file or NULL if no report should be produced. Defaults to {user_home}/ISAnalytics_reports.

Details

This particular test calculates for each column in the key

- The zscore of the values
- The tstudent of the values
- The the distribution of the tstudent values

Optionally the test can be performed for each pool and a normality test can be run prior the actual calculations. Samples are flagged if this condition is respected:

• tdist < outlier_p_value_threshold & zscore < 0

If the key contains more than one column an additional flag logic can be specified for combining the results. Example: let's suppose the key contains the names of two columns, X and Y key = c("X", "Y") if we specify the the argument flag_logic = "AND" then the reads will be flagged based on this global condition: (tdist_X < outlier_p_value_threshold & zscore_X < 0) AND (tdist_Y < outlier_p_value_threshold & zscore_Y < 0)

The user can specify one or more logical operators that will be applied in sequence.

Value

A data frame of metadata with the column to_remove

See Also

```
Other Outlier tests: available_outlier_tests()
```

```
data("association_file", package = "ISAnalytics")
flagged <- outliers_by_pool_fragments(association_file,
    report_path = NULL
)
head(flagged)</pre>
```

outlier_filter 47

outlier_filter

Filter out outliers in metadata, identified by the chosen outlier test.

Description

[Experimental] Filter out outliers in metadata.

Usage

```
outlier_filter(
  metadata,
  outlier_test = "outliers_by_pool_fragments",
  negate = FALSE,
   ...
)
```

Arguments

metadata	The metadata data frame
outlier_test	A string representing a function name. The name must be one of the available outlier tests, see available_outlier_tests.
negate	If TRUE will return only the metadata that was flagged to be removed. If FALSE will return only the metadata that wasn't flagged to be removed.
	Additional named arguments passed to outliers_test

Value

A data frame of metadata which has less or the same amount of rows

```
data("association_file", package = "ISAnalytics")
filtered_af <- outlier_filter(association_file,
    key = "BARCODE_MUX",
    report_path = NULL
)
head(filtered_af)</pre>
```

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proto_oncogenes

Data frames for proto-oncogenes (human and mouse) amd tumorsuppressor genes from UniProt.

Description

The file is simply a result of a research with the keywords "proto-oncogenes" and "tumor suppressor" for the target genomes on UniProt database.

Usage

```
data("proto_oncogenes")
data("tumor_suppressors")
```

Format

An object of class tbl_df (inherits from tbl, data.frame) with 569 rows and 13 columns. An object of class tbl_df (inherits from tbl, data.frame) with 523 rows and 13 columns.

Functions

• tumor_suppressors: Data frame for tumor suppressor genes

purity_filter

Filter integration sites based on purity.

Description

[Experimental] Filter that targets possible contamination between cell lines based on a numeric quantification (likely abundance or sequence count).

Usage

```
purity_filter(
    x,
    lineages = blood_lineages_default(),
    aggregation_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"),
    group_key = c("CellMarker", "Tissue"),
    selected_groups = NULL,
    join_on = "CellMarker",
    min_value = 3,
    impurity_threshold = 10,
    by_timepoint = TRUE,
    timepoint_column = "TimePoint",
    value_column = "seqCount_sum"
)
```

purity_filter 49

Arguments

x An aggregated integration matrix, obtained via aggregate_values_by_key()

lineages A data frame containing cell lineages information

aggregation_key

The key used for aggregating x

group_key A character vector of column names for re-aggregation. Column names must be

either in x or in lineages. See details.

selected_groups

Either NULL, a character vector or a data frame for group selection. See details.

join_on Common columns to perform a join operation on

min_value A minimum value to filter the input matrix. Integrations with a value strictly

lower than min_value are excluded (dropped) from the output.

impurity_threshold

The ratio threshold for impurity in groups

by_timepoint Should filtering be applied on each time point? If FALSE, all time points are

merged together

timepoint_column

Column in x containing the time point

value_column Column in x containing the numeric quantification of interest

Details

Setting input arguments:

The input matrix can be re-aggregated with the provided group_key argument. This key contains the names of the columns to group on (besides the columns holding genomic coordinates of the integration sites) and must be contained in at least one of x or lineages data frames. If the key is not found only in x, then a join operation with the lineages data frame is performed on the common column(s) join_on.

Group selection:

It is possible for the user to specify on which groups the logic of the filter should be applied to. For example: if we have group_key = c("HematoLineage") and we set selected_groups = c("CD34", "Myeloid", "Lymphoid") it means that a single integration will be evaluated for the filter only for groups that have the values of "CD34", "Myeloid" and "Lymphoid" in the "HematoLineage" column. If the same integration is present in other groups it is kept as it is. selected_groups can be set to NULL if we want the logic to apply to every group present in the data frame, it can be set as a simple character vector as the example above if the group key has length 1 (and there is no need to filter on time point). If the group key is longer than 1 then the filter is applied only on the first element of the key.

If a more refined selection on groups is needed, a data frame can be provided instead:

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```
"CD14", "PB"
```

Columns in the data frame should be the same as group key (plus, eventually, the time point column). In this example only those groups identified by the rows in the provided data frame are processed.

Value

A data frame

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), sample_statistics(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
filtered_by_purity <- purity_filter(
    x = aggreg,
    value_column = "seqCount_sum"
)
head(filtered_by_purity)</pre>
```

 ${\tt quantification_types} \quad \textit{Possible choices for the } {\tt quantification_type} \ \textit{parameter}.$

Description

These are all the possible values for the quantification_type parameter in import_parallel_vispa2Matrices_interac and import_parallel_vispa2Matrices_auto.

Usage

```
quantification_types()
```

Details

The possible values are:

- fragmentEstimate
- seqCount
- · barcodeCount
- · cellCount
- ShsCount

Value

A vector of characters for quantification types

See Also

```
import_parallel_Vispa2Matrices_interactive, import_parallel_Vispa2Matrices_auto
Other Import functions helpers: annotation_issues(), matching_options()
```

Examples

```
quant_types <- quantification_types()</pre>
```

```
realign_after_collisions
```

Re-aligns matrices of other quantification types based on the processed sequence count matrix.

Description

[Stable] This function should be used to keep data consistent among the same analysis: if for some reason you removed the collisions by passing only the sequence count matrix to remove_collisions(), you should call this function afterwards, providing a list of other quantification matrices. NOTE: if you provided a list of several quantification types to remove_collisions() before, there is no need to call this function.

Usage

```
realign_after_collisions(sc_matrix, other_matrices)
```

Arguments

```
sc_matrix The sequence count matrix already processed for collisions via remove_collisions()
other_matrices A named list of matrices to re-align. Names in the list must be quantification
types (quantification_types()) except "seqCount".
```

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Details

For more details on how to use collision removal functionality: vignette("collision_removal", package = "ISAnalytics")

Value

A named list with re-aligned matrices

See Also

```
remove_collisions
Other Collision removal: remove_collisions()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
separated <- separate_quant_matrices(
    integration_matrices)
)
no_coll <- remove_collisions(
    x = separated$seqCount,
    association_file = association_file,
    quant_cols = c(seqCount = "Value"),
    report_path = NULL
)
realigned <- realign_after_collisions(
    sc_matrix = no_coll,
    other_matrices = list(fragmentEstimate = separated$fragmentEstimate)
)
realigned</pre>
```

reduced_AF_columns

Names of the columns of the association file to consider for Vispa2 launch.

Description

Selection of column names from the association file to be considered for Vispa2 launch. NOTE: the TagID column appears only once but needs to be repeated twice for generating the launch file. Use the appropriate function to generate the file automatically.

Usage

```
reduced_AF_columns()
```

Value

A character vector

refGenes_hg19 53

Examples

```
reduced_AF_columns()
```

refGenes_hg19

Gene annotation files for hg19, mm9 and mm10.

Description

This file was obtained following this steps:

- 1. Download from http://hgdownload.soe.ucsc.edu/goldenPath/hg19/database/ the refGene.sql, knownGene.sql, knownToRefSeq.sql, kgXref.sql tables
- 2. Import everything it in mysql
- 3. Generate views for annotation:

```
SELECT kg.`chrom`, min(kg.cdsStart) as CDS_minStart,
max(kg.`cdsEnd`) as CDS_maxEnd, k2a.geneSymbol,
kg.`strand` as GeneStrand, min(kg.txStart) as TSS_minStart,
max(kg.txEnd) as TSS_maxStart,
kg.proteinID as ProteinID, k2a.protAcc as ProteinAcc, k2a.spDisplayID
FROM `knownGene` AS kg JOIN kgXref AS k2a
ON BINARY kg.name = k2a.kgID COLLATE latin1_bin
-- latin1_swedish_ci
-- WHERE k2a.spDisplayID IS NOT NULL and (k2a.`geneSymbol` LIKE 'Tcra%' or
k2a.`geneSymbol` LIKE 'TCRA%')
WHERE (k2a.spDisplayID IS NOT NULL or k2a.spDisplayID NOT LIKE '')
and k2a.`geneSymbol` LIKE 'Tcra%'
group by kg.`chrom`, k2a.geneSymbol
ORDER BY kg.chrom ASC , kg.txStart ASC
```

Usage

```
data("refGenes_hg19")
data("refGenes_mm9")
```

Format

An object of class tbl_df (inherits from tbl, data.frame) with 27275 rows and 12 columns. An object of class tbl_df (inherits from tbl, data.frame) with 24487 rows and 12 columns.

Functions

• refGenes_mm9: Data frame for murine mm9 genome

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refGene_table_cols

Required columns for refGene file.

Description

Required columns for refGene file.

Usage

```
refGene_table_cols()
```

Value

Character vector of column names

Examples

```
refGene_table_cols()
```

remove_collisions

Identifies and removes collisions.

Description

[Stable] A collision is an integration (aka a unique combination of chr, integration_locus and strand) which is observed in more than one independent sample (a unique pair of ProjectID and SubjectID). The function tries to decide to which subject an integration should be assigned to and, if no decision can be taken, the integration is completely removed from the data frame. For more details refer to the vignette "Collision removal functionality": vignette("collision_removal", package = "ISAnalytics")

Usage

```
remove_collisions(
    x,
    association_file,
    date_col = "SequencingDate",
    reads_ratio = 10,
    quant_cols = c(seqCount = "seqCount", fragmentEstimate = "fragmentEstimate"),
    report_path = default_report_path(),
    max_workers = NULL
)
```

sample_statistics 55

Arguments

X	Either a multi-quantification matrix or a named list of matrices (names must be quantification types)
association_fi	le
	The association file imported via import_association_file()
date_col	The date column that should be considered. Must be one value in date_columns_coll()
reads_ratio	A single numeric value that represents the ratio that has to be considered when deciding between seqCount value.
quant_cols	A named character vector where names are quantification types and values are the names of the corresponding columns. The quantification seqCount MUST be included in the vector.
report_path	The path where the report file should be saved. Can be a folder, a file or NULL if no report should be produced. Defaults to {user_home}/ISAnalytics_reports.
max_workers	Maximum number of parallel workers to distribute the workload. If NULL (default) produces the maximum amount of workers allowed, a numeric value is requested otherwise. WARNING: a higher number of workers speeds up com-

putation at the cost of memory consumption! Tune this parameter accordingly.

Value

Either a multi-quantification matrix or a list of data frames

See Also

```
date_columns_coll
Other Collision removal: realign_after_collisions()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
no_coll <- remove_collisions(
    x = integration_matrices,
    association_file = association_file,
    report_path = NULL
)
head(no_coll)</pre>
```

sample_statistics

Computes user specified functions on numerical columns and updates the metadata data frame accordingly.

56 sample_statistics

Description

[Experimental] The function operates on a data frame by grouping the content by the sample key and computing every function specified on every column in the value_columns parameter. After that the metadata data frame is updated by including the computed results as columns for the corresponding key. For this reason it's required that both x and metadata have the same sample key, and it's particularly important if the user is working with previously aggregated data. For example:

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
aggreg_meta <- aggregate_metadata(association_file = association_file)
sample_stats <- sample_statistics(x = aggreg,
    metadata = aggreg_meta,
    value_columns = c("seqCount", "fragmentEstimate"),
sample_key = c("SubjectID", "CellMarker", "Tissue", "TimePoint"))</pre>
```

Usage

```
sample_statistics(
   x,
   metadata,
   sample_key = "CompleteAmplificationID",
   value_columns = "Value",
   functions = default_stats(),
   add_integrations_count = TRUE
)
```

Arguments

x A data frame

sample_key Character vector representing the key for identifying a sample

value_columns The name of the columns to be computed, must be numeric or integer

functions A named list of function or purrr-style lambdas

add_integrations_count

Add the count of distinct integration sites for each group? Can be computed only if x contains the mandatory columns chr, integration_locus, strand

Value

A list with modified x and metadata data frames

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), separate_quant_matrices(), threshold_filter(), top_integrations()
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
stats <- sample_statistics(
    x = integration_matrices,
    metadata = association_file,
    value_columns = c("seqCount", "fragmentEstimate")
)
stats</pre>
```

separate_quant_matrices

Separate a multiple-quantification matrix into single quantification matrices.

Description

[Stable] The function separates a single multi-quantification integration matrix, obtained via comparison_matrix, into single quantification matrices as a named list of tibbles.

Usage

```
separate_quant_matrices(
    x,
    fragmentEstimate = "fragmentEstimate",
    seqCount = "seqCount",
    barcodeCount = "barcodeCount",
    cellCount = "cellCount",
    ShsCount = "ShsCount",
    key = c(mandatory_IS_vars(), annotation_IS_vars(), "CompleteAmplificationID")
)
```

Arguments

x Single integration matrix with multiple quantification value columns, likely obtained via comparison_matrix.

fragmentEstimate

Name of the fragment estimate values column in input

seqCount Name of the sequence count values column in input barcodeCount Name of the barcode count values column in input

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cellCount	Name of the cell count values column in input
ShsCount	Name of the shs count values column in input
key	Key columns to perform the joining operation

Value

A named list of tibbles, where names are quantification types

See Also

```
quantification_types
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(),
cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(),
```

Examples

```
data("integration_matrices", package = "ISAnalytics")
separated <- separate_quant_matrices(
    integration_matrices
)
separated</pre>
```

threshold_filter(), top_integrations()

sharing_heatmap

Plot IS sharing heatmaps.

Description

[Experimental] Displays the IS sharing calculated via is_sharing as heatmaps.

Usage

```
sharing_heatmap(
    sharing_df,
    show_on_x = "g1",
    show_on_y = "g2",
    absolute_sharing_col = "shared",
    title_annot = NULL,
    plot_relative_sharing = TRUE,
    rel_sharing_col = c("on_g1", "on_union"),
    show_perc_symbol_rel = TRUE,
    interactive = FALSE
)
```

sharing_heatmap 59

Arguments

```
The data frame containing the IS sharing data
sharing_df
                  Name of the column to plot on the x axis
show_on_x
                  Name of the column to plot on the y axis
show_on_y
absolute_sharing_col
                  Name of the column that contains the absolute values of IS sharing
title_annot
                  Additional text to display in the title
plot_relative_sharing
                  Logical. Compute heatmaps also for relative sharing?
rel_sharing_col
                  Names of the columns to consider as relative sharing. The function is going to
                  plot one heatmap per column in this argument.
show_perc_symbol_rel
                  Logical. Only relevant if plot_relative_sharing is set to TRUE, should the
                  percentage symbol be displayed in relative heatmaps?
interactive
                  Logical. Requires the package plotly is required for this functionality. Returns
                  the heatmaps as interactive HTML widgets.
```

Value

A list of plots or widgets

See Also

```
is_sharing
```

```
Other Plotting functions: CIS_volcano_plot(), HSC_population_plot(), circos_genomic_density(), integration_alluvial_plot(), sharing_venn(), top_abund_tableGrob()
```

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)
sharing <- is_sharing(aggreg,
    minimal = FALSE,
    include_self_comp = TRUE
)
sharing_heatmaps <- sharing_heatmap(sharing_df = sharing)
sharing_heatmaps$absolute
sharing_heatmaps$on_g1
sharing_heatmaps$on_union</pre>
```

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sharing_venn	Produce tables to plot sharing venn or euler diagrams.	

Description

[Experimental] This function processes a sharing data frame obtained via is_sharing() with the option table_for_venn = TRUE to obtain a list of objects that can be plotted as venn or euler diagrams.

Usage

```
sharing_venn(sharing_df, row_range = NULL, euler = TRUE)
```

Arguments

sharing_df The sharing data frame

row_range Either NULL or a numeric vector of row indexes (e.g. c(1,4,5) will produce

tables only for rows 1, 4 and 5)

euler If TRUE will produce tables for euler diagrams, otherwise will produce tables for

venn diagrams

Details

The functions requires the package eulerr. Each row of the input data frame is representable as a venn/euler diagram. The function allows to specify a range of row indexes to obtain a list of plottable objects all at once, leave it to NULL to process all rows.

To actually plot the data it is sufficient to call the function plot() and specify optional customization arguments. See <u>eulerr docs</u> for more detail on this.

Value

A list of data frames

See Also

```
Other Plotting functions: CIS_volcano_plot(), HSC_population_plot(), circos_genomic_density(), integration_alluvial_plot(), sharing_heatmap(), top_abund_tableGrob()
```

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(
    x = integration_matrices,
    association_file = association_file,
    value_cols = c("seqCount", "fragmentEstimate")
)</pre>
```

threshold_filter 61

```
sharing <- is_sharing(aggreg, n_comp = 3, table_for_venn = TRUE)
venn_tbls <- sharing_venn(sharing, row_range = 1:3, euler = FALSE)
venn_tbls
plot(venn_tbls[[1]])</pre>
```

threshold_filter

Filter data frames with custom predicates

Description

[Experimental] Filter a single data frame or a list of data frames with custom predicates assembled from the function parameters.

Usage

```
threshold_filter(x, threshold, cols_to_compare = "Value", comparators = ">")
```

Arguments

x A data frame or a list of data frames

threshold A numeric/integer vector or a named list of numeric/integer vectors

cols_to_compare

A character vector or a named list of character vectors

comparators A character vector or a named list of character vectors. Must be one of the

allowed values between c("<", ">", "==", "!=", ">=", "<=")

Details

A single data frame as input:

If the user chooses to operate on a single data frame, the other parameters should only be vectors: numeric vector for threshold and character vectors for both cols_to_compare and comparators. A filtering condition is obtained by combining element by element cols_to_compare + comparators + threshold (similarly to the paste function). For example:

```
threshold = c(20, 35, 50)

cols\_to\_compare = c("a", "b", "c")

comparators = "<"
```

given these vectors, the input data frame will be filtered by checking which values in column "a" are less than 20 **AND** which values in column "b" are less than 35 **AND** which values in column "c" are less than 50. Things the user should keep in mind are:

- The vectors of length 1 are going to be recycled if one or more parameters are longer (in the example, the comparators value)
- If vectors are not of length 1 they must have the same length
- Columns to compare, of course, need to be included in the input data frame and need to be numeric/integer
- The filtering will perform a logical "AND" on all the conditions, only rows that satisfy ALL the conditions are preserved

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A list of data frames as input:

The input for the function may also be a list of data frames, either named or unnamed.

Unnamed list:

If the input is a simple unnamed list, the other parameters should be simple vectors (as for data frames). All the predicates will simply be applied to every data frame in the list: this is useful if it's desirable to filter for the same conditions different data frames that have the same structure but different data.

Named list:

It is also possible to filter different data frames with different sets of conditions. Besides having the possibility of defining the other parameters as simple vector, which has the same results as operating on an unnamed list, the user can define the parameters as named lists containing vectors. For example:

```
example_df <- tibble::tibble(a = c(20, 30, 40),
                              b = c(40, 50, 60),
                              c = c("a", "b", "c"),
                              d = c(3L, 4L, 5L)
example_list <- list(first = example_df,
                      second = example_df,
                      third = example_df)
print(example_list)
## $first
## # A tibble: 3 × 4
##
         а
               b c
##
     <dbl> <dbl> <chr> <int>
## 1
        20
              40 a
                            3
## 2
        30
              50 b
                            4
## 3
        40
                            5
              60 c
##
## $second
## # A tibble: 3 × 4
##
         а
               b c
##
     <dbl> <dbl> <chr> <int>
## 1
        20
              40 a
                            3
## 2
              50 b
                            4
        30
## 3
        40
              60 c
##
## $third
## # A tibble: 3 × 4
##
               b c
         а
     <dbl> <dbl> <chr> <int>
##
## 1
        20
              40 a
                            3
## 2
        30
              50 b
                            4
## 3
        40
              60 c
filtered <- threshold_filter(example_list,</pre>
threshold = list(first = c(20, 60),
third = c(25)),
cols_to_compare = list(first = c("a", "b"),
```

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```
third = c("a"),
comparators = list(first = c(">", "<"),</pre>
third = c(">="))
print(filtered)
## $first
## # A tibble: 1 × 4
        а
              bс
     <dbl> <dbl> <chr> <int>
## 1
       30
              50 b
##
## $second
## # A tibble: 3 × 4
##
              b c
        а
     <dbl> <dbl> <chr> <int>
##
## 1
       20 40 a
## 2
        30
              50 b
## 3
       40
             60 c
##
## $third
## # A tibble: 2 × 4
##
        a
              b c
##
     <dbl> <dbl> <chr> <int>
## 1
        30
              50 b
## 2
        40
              60 c
```

The above signature will roughly be translated as:

- Filter the element "first" in the list by checking that values in column "a" are bigger than 20 AND values in column "b" are less than 60
- Don't apply any filter to the element "second" (returns the data frame as is)
- Filter the element "third" by checking that values in column "a" are equal or bigger than 25.

It is also possible to use some parameters as vectors and some as lists: vectors will be recycled for every element filtered.

```
filtered <- threshold_filter(example_list,
threshold = list(first = c(20, 60),
third = c(25, 65)),
cols_to_compare = c("a", "b"),
comparators = list(first = c(">", "<"),
third = c(">=", "<=")))</pre>
```

In this example, different threshold and comparators will be applied to the same columns in all data frames.

Things the user should keep in mind are:

- Names for the list parameters must be the same names in the input list
- Only elements explicited in list parameters as names will be filtered
- · Lengths of both vectors and lists must be consistent

Value

A data frame or a list of data frames

See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), top_integrations()
```

Examples

```
example_df <- tibble::tibble(</pre>
   a = c(20, 30, 40),
   b = c(40, 50, 60),
   c = c("a", "b", "c"),
   d = c(3L, 4L, 5L)
)
example_list <- list(</pre>
   first = example_df,
    second = example_df,
    third = example_df
)
filtered <- threshold_filter(example_list,</pre>
    threshold = list(
        first = c(20, 60),
        third = c(25)
   ),
    cols_to_compare = list(
        first = c("a", "b"),
        third = c("a")
    ),
    comparators = list(
        first = c(">", "<"),
        third = c(">=")
filtered
```

top_abund_tableGrob Summary top abundant tableGrobs for plots.

Description

Produce summary tableGrobs as R graphics. For this functionality the suggested package gridExtra is required. To visualize the resulting object:

```
gridExtra::grid.arrange(tableGrob)
```

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Usage

```
top_abund_tableGrob(
   df,
   id_cols = mandatory_IS_vars(),
   quant_col = "fragmentEstimate_sum_PercAbundance",
   by = "TimePoint",
   alluvial_plot = NULL,
   top_n = 10,
   tbl_cols = "GeneName",
   include_id_cols = FALSE,
   digits = 2,
   perc_symbol = TRUE
)
```

Arguments

df	A data frame
id_cols	Character vector of id column names. To plot after alluvial, these columns must be the same as the alluvia argument of integration_alluvial_plot.
quant_col	Column name holding the quantification value. To plot after alluvial, these columns must be the same as the plot_y argument of integration_alluvial_plot.
by	The column name to subdivide tables for. The function will produce one table for each distinct value in by. To plot after alluvial, these columns must be the same as the plot_x argument of integration_alluvial_plot.
alluvial_plot	Either NULL or an alluvial plot for color mapping between values of y.
top_n	Integer. How many rows should the table contain at most?
tbl_cols	Table columns to show in the final output besides quant_col.
include_id_cols	
	Logical. Include id_cols in the output?
digits	Integer. Digits to show for the quantification column
perc_symbol	Logical. Show percentage symbol in the quantification column?

Value

A tableGrob object

See Also

```
Other Plotting functions: CIS_volcano_plot(), HSC_population_plot(), circos_genomic_density(), integration_alluvial_plot(), sharing_heatmap(), sharing_venn()
```

```
data("integration_matrices", package = "ISAnalytics")
data("association_file", package = "ISAnalytics")
aggreg <- aggregate_values_by_key(</pre>
```

top_integrations

```
x = integration_matrices,
   association_file = association_file,
   value_cols = c("seqCount", "fragmentEstimate")
)
abund <- compute_abundance(x = aggreg)
grob <- top_abund_tableGrob(abund)
gridExtra::grid.arrange(grob)</pre>
```

top_integrations

Sorts and keeps the top n integration sites based on the values in a given column.

Description

[Experimental] The input data frame will be sorted by the highest values in the columns specified and the top n rows will be returned as output. The user can choose to keep additional columns in the output by passing a vector of column names or passing 2 "shortcuts":

- keep = "everything" keeps all columns in the original data frame
- keep = "nothing" only keeps the mandatory columns (mandatory_IS_vars()) plus the columns in the columns parameter.

Usage

```
top_integrations(
    x,
    n = 20,
    columns = "fragmentEstimate_sum_RelAbundance",
    keep = "everything",
    key = NULL
)
```

Arguments

Х	An integration matrix (data frame containing mandatory_IS_vars())
n	How many integrations should be sliced (in total or for each group)? Must be numeric or integer and greater than 0
columns	Columns to use for the sorting. If more than a column is supplied primary ordering is done on the first column, secondary ordering on all other columns
keep	Names of the columns to keep besides mandatory_IS_vars() and columns
key	Either NULL or a character vector of column names to group by. If not NULL the input will be grouped and the top fraction will be extracted from each group.

Value

Either a data frame with at most n rows or a data frames with at most n*(number of groups) rows.

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See Also

```
Other Analysis functions: CIS_grubbs(), comparison_matrix(), compute_abundance(), cumulative_count_union(), cumulative_is(), is_sharing(), iss_source(), purity_filter(), sample_statistics(), separate_quant_matrices(), threshold_filter()
```

Examples

```
smpl <- tibble::tibble(</pre>
    chr = c("1", "2", "3", "4", "5", "6"),
    integration_locus = c(14536, 14544, 14512, 14236, 14522, 14566),
    strand = c("+", "+", "-", "+", "-", "+"),
    CompleteAmplificationID = c("ID1", "ID2", "ID1", "ID1", "ID3", "ID2"),
    Value = c(3, 10, 40, 2, 15, 150),
    Value2 = c(456, 87, 87, 9, 64, 96),
    Value3 = c("a", "b", "c", "d", "e", "f")
)
top <- top_integrations(smpl,</pre>
   n = 3,
   columns = c("Value", "Value2"),
   keep = "nothing"
top_key <- top_integrations(smpl,</pre>
   n = 3,
   columns = "Value",
   keep = "Value2",
   key = "CompleteAmplificationID"
)
```

unzip_file_system

A utility function to unzip and use example file systems included in the package

Description

This utility function is a simple shortcut to create a temporary directory, unzip and reference the examples file systems included in the package for testing purposes.

Usage

```
unzip_file_system(zipfile, name)
```

Arguments

zipfile The zipped file to decompress

name The name of the folder in the zipped archive ("fs" or "fserr")

Value

A path to reference

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See Also

 $Other\ Utility\ functions:\ as_sparse_matrix(),\ generate_Vispa2_launch_AF(),\ generate_blank_association_file()$

```
root_pth <- system.file("extdata", "fs.zip", package = "ISAnalytics")
root <- unzip_file_system(root_pth, "fs")</pre>
```

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