

Package ‘YAPSA’

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Description This package provides functions and routines for supervised analyses of mutational signatures (i.e., the signatures have to be known, cf. L. Alexandrov et al., Nature 2013 and L. Alexandrov et al., BioRxiv 2018). In particular, the family of functions LCD (LCD = linear combination decomposition) can use optimal signature-specific cutoffs which takes care of different detectability of the different signatures. Moreover, the package provides different sets of mutational signatures, including the COSMIC and PCAWG SNV signatures and the PCAWG Indel signatures; the latter inferring that with YAPSA, the concept of supervised analysis of mutational signatures is extended to Indel signatures. YAPSA also provides confidence intervals as computed by profile likelihoods and can perform signature analysis on a stratified mutational catalogue (SMC = stratify mutational catalogue) in order to analyze enrichment and depletion patterns for the signatures in different strata.

License GPL-3

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add_annotation	<i>Add information to an annotation data structure</i>
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Description

Function to iteratively add information to an annotation data structure as needed for [HeatmapAnnotation](#) and especially for [annotation_exposures_barplot](#)

Usage

```
add_annotation(
  in_annotation_col,
  in_annotation_df,
  in_attribution_vector,
  in_colour_vector,
  in_name
)
```

Arguments

in_annotation_col	List, every element of which refers to one layer of annotation List elements are structures corresponding to named colour vectors
in_annotation_df	Data frame, every column of which corresponds to a layer of annotation. It has as many rows as there are samples, every entry in a row corresponding to the attribute the samples has for the corresponding layer of annotation. The factor levels of a column of in_annotation_df correspond to the names of the corresponding element in in_annotation_col
in_attribution_vector	A vector which is going to be cbinded to in_annotatiin_df, carrying the annotation information of the new layer to be added
in_colour_vector	Named vector of colours to be attributed to the new annotation
in_name	Name of the new layer of annotation

Value

A list with entries

- `annotation_col`: A list as in `in_annotation_col` but with one additional layer of annotation
- `annotation_df`: A data frame as in `in_annotation_df` but with one additional layer of annotation

Examples

NULL

`add_as_fist_to_list` *Add an element as first entry to a list*

Description

Works for all types of lists and inputs

Usage

```
add_as_fist_to_list(in_list, in_element)
```

Arguments

<code>in_list</code>	List to which an element is to be added
<code>in_element</code>	Element to be added

Value

List with input element as first entry.

Examples

NULL

`aggregate_exposures_by_category`
Aggregate exposures by category

Description

If a valid category (i.e. it matches to a category specified in `in_sig_ind_df`) is supplied, then the exposures are aggregated over this category.

Usage

```
aggregate_exposures_by_category(in_exposures_df, in_sig_ind_df, in_category)
```

Arguments

in_exposures_df	Input data frame of exposures.
in_sig_ind_df	Input data frame of meta information on the signatures. Has to match the signatures in in_exposures_df
in_category	Category to be aggregated over

Value

A list with entries:

- exposures: The exposures H, a numeric data frame with l rows and m columns, l being the number of aggregated signatures and m being the number of samples
- norm_exposures: The normalized exposures H, a numeric data frame with l rows and m columns, l being the number of aggregated signatures and m being the number of samples
- out_sig_ind_df: Data frame of the type signature_indices_df, i.e. indicating name, function and meta-information of the aggregated signatures..

See Also

[LCD_complex_cutoff](#)

Examples

NULL

annotate_intermut_dist_cohort

Annotate the intermutation distance of variants cohort-wide

Description

The function annotates the intermutational distance to a cohort wide data frame by applying [annotate_intermut_dist_PID](#) to every PID-specific subfraction of the cohort wide data. Note that [annotate_intermut_dist_PID](#) calls [rainfallTransform](#). If the PID information is missing, [annotate_intermut_dist_PID](#) is called directly for the whole input.

Usage

```
annotate_intermut_dist_cohort(
  in_dat,
  in_CHROM.field = "CHROM",
  in_POS.field = "POS",
  in_PID.field = NULL,
  in_mode = "min",
  in_verbose = FALSE
)
```

Arguments

<code>in_dat</code>	VRanges object, VRangesList, data frame or list of data frames which carries (at least) one column for the chromosome and one column for the position. Optionally, a column to specify the PID can be provided.
<code>in_CHROM.field</code>	String indicating which column of <code>in_df</code> carries the chromosome information
<code>in_POS.field</code>	String indicating which column of <code>in_df</code> carries the position information
<code>in_PID.field</code>	String indicating which column of <code>in_df</code> carries the PID information
<code>in_mode</code>	String passed through <code>annotate_intermut_dist_PID</code> to <code>rainfallTransform</code> indicating which method to choose for the computation of the intermutational distance.
<code>in_verbose</code>	Whether verbose or not.

Value

VRanges object, VRangesList, data frame or list of data frames identical to `in_df` (reordered by `in_PID.field`), but with the intermutation distance annotated as an additional column on the right named `dist`.

See Also

[annotate_intermut_dist_PID](#)

[rainfallTransform](#)

Examples

```
test_df <- data.frame(CHROM=c(1,1,1,2,2,2,3,3,3,4,4,4,5,5),
                    POS=c(1,2,4,4,6,9,1,4,8,10,20,40,100,200),
                    REF=c("C","C","C","T","T","T","A",
                          "A","A","G","G","G","N","A"),
                    ALT=c("A","G","T","A","C","G","C",
                          "G","T","A","C","T","A","N"),
                    PID=c(1,1,1,2,2,2,1,1,2,2,2,1,1,2))
test_df <- test_df[order(test_df$PID, test_df$CHROM, test_df$POS),]
min_dist_df <-
  annotate_intermut_dist_cohort(test_df, in_CHROM.field="CHROM",
                              in_POS.field="POS", in_PID.field="PID",
                              in_mode="min")
max_dist_df <-
  annotate_intermut_dist_cohort(test_df, in_CHROM.field="CHROM",
                              in_POS.field="POS", in_PID.field="PID",
                              in_mode="max")

min_dist_df
max_dist_df
```



```

max_dist_df <- annotate_intermut_dist_PID(test_df, in_CHROM.field="CHROM",
                                         in_POS.field="POS",
                                         in_mode="max")

min_dist_df
max_dist_df

```

annotation_exposures_barplot

Plot the exposures of a cohort with different layers of annotation

Description

The exposures H, determined by NMF or by LCD, are displayed as a stacked barplot by calling [Heatmap](#). The x-axis displays the PIDs (patient identifier or sample), the y-axis the counts attributed to the different signatures with their respective colours per PID. It is analogous to [plot_exposures](#). As many layers of information as desired can be added via an annotation data frame. The annotation data is handled in a way similar to [annotation_heatmap_exposures](#). This function calls:

- [rowAnnotation](#),
- [HeatmapAnnotation](#) and
- [Heatmap](#)

Usage

```

annotation_exposures_barplot(
  in_exposures_df,
  in_signatures_ind_df,
  in_subgroups_df,
  in_annotation_df = NULL,
  in_annotation_col = NULL,
  ylab = NULL,
  title = "",
  in_labels = FALSE,
  in_barplot_borders = TRUE,
  in_column_anno_borders = FALSE,
  in_annotation_legend_side = "right",
  in_padding = unit(c(2, 20, 2, 2), "mm"),
  in_annotation = NULL
)

```

Arguments

- `in_exposures_df`
Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).
- `in_signatures_ind_df`
A data frame containing meta information about the signatures
- `in_subgroups_df`
A data frame indicating which PID (patient or sample identifier) belongs to which subgroup

<code>in_annotation_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup for all layers of annotation
<code>in_annotation_col</code>	A list indicating colour attributions for all layers of annotation
<code>ylab</code>	String indicating the column name in <code>in_subgroups_df</code> to take the subgroup information from.
<code>title</code>	Title for the plot to be created.
<code>in_labels</code>	Whether or not to show the names of the samples.
<code>in_barplot_borders</code>	Whether or not to show border lines in barplot
<code>in_column_anno_borders</code>	Whether or not to draw separating lines between the fields in the annotation
<code>in_annotation_legend_side</code>	Where to put the legends of the annotation df, default is right.
<code>in_padding</code>	Parameter passed on to function draw
<code>in_annotation</code>	A full annotation object may also be provided by the educated user.

Details

It might be necessary to install the newest version of the development branch of the packages **circlize** and **ComplexHeatmap** by Zuguang Gu: `devtools::install_github("jokergoo/circlize")` and `devtools::install_github("jokergoo/ComplexHeatmap")`

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Value

The function doesn't return any value.

See Also

[HeatmapAnnotation](#)
[Heatmap](#)
[decorate_heatmap_body](#)
[annotation_heatmap_exposures](#)
[plot_exposures](#)

Examples

NULL

```
annotation_exposures_list_barplot
```

Plot the exposures of a cohort with different layers of annotation for SNV and INDEL signatures

Description

The exposures H, determined by NMF or by LCD, are displayed as a stacked barplot by calling [Heatmap](#). The x-axis displays the PIDs (patient identifier or sample), the y-axis the counts attributed to the different signatures with their respective colours per PID. It is analogous to [plot_exposures](#). As many layers of information as desired can be added via an annotation data frame. The annotation data is handled in a way similar to [annotation_heatmap_exposures](#). In comparison to [annotation_exposures_barplot](#) allows this function to deal with a list of different signature and mutation types. This function calls:

- [rowAnnotation](#),
- [HeatmapAnnotation](#) and
- [Heatmap](#)

Usage

```
annotation_exposures_list_barplot(
  in_exposures_list,
  in_signatures_ind_list,
  in_subgroups_list,
  in_annotation_list,
  ylab = NULL,
  title = "",
  in_indel_sigs = FALSE,
  in_labels = FALSE,
  in_barplot_borders = TRUE,
  in_column_anno_borders = FALSE,
  in_annotation_legend_side = "right",
  in_padding = unit(c(2, 20, 2, 2), "mm"),
  in_annotation = NULL
)
```

Arguments

`in_exposures_list`

A list of numerical data frame encoding the exposures H of different signature types, i.e. which signature contributes how much to which PID (patient identifier or sample).

`in_signatures_ind_list`

A list of data frame containing meta information about the each signature type individually

`in_subgroups_list`

A list of data frame indicating of each signature type which PID (patient or sample identifier) belongs to which subgroup

<code>in_annotation_list</code>	A list data frame indicating which PID (patient or sample identifier) belongs to which subgroup for all layers of annotation and a list indicating colour attributions for all layers of annotation for each signature type individually
<code>ylab</code>	String indicating the column name in <code>in_subgroups_df</code> to take the subgroup information from.
<code>title</code>	Title for the plot to be created.
<code>in_indel_sigs</code>	Tag which is default FALSE when whole genome data are analysed the tag will be TRUE
<code>in_labels</code>	Whether or not to show the names of the samples.
<code>in_barplot_borders</code>	Whether or not to show border lines in barplot
<code>in_column_anno_borders</code>	Whether or not to draw separating lines between the fields in the annotation
<code>in_annotation_legend_side</code>	Where to put the legends of the annotation df, default is right.
<code>in_padding</code>	Parameter passed on to function draw
<code>in_annotation</code>	A full annotation object may also be provided by the educated user.

Details

It might be necessary to install the newest version of the development branch of the packages **circclize** and **ComplexHeatmap** by Zuguang Gu: `devtools::install_github("jokergoo/circlize")` and `devtools::install_github("jokergoo/ComplexHeatmap")`

It might be necessary to install the newest version of the development branch of the packages **circclize** and **ComplexHeatmap** by Zuguang Gu: `devtools::install_github("jokergoo/circlize")` and `devtools::install_github("jokergoo/ComplexHeatmap")`

Value

The function doesn't return any value.

See Also

[HeatmapAnnotation](#)
[Heatmap](#)
[decorate_heatmap_body](#)
[annotation_heatmap_exposures](#)
[plot_exposures](#)

Examples

NULL

 annotation_heatmap_exposures

Heatmap to cluster the PIDs on their signature exposures (Complex-Heatmap)

Description

The PIDs are clustered according to their signature exposures. The procedure is analogous to [complex_heatmap_exposures](#), but enabling more than one annotation row for the PIDs. This function calls:

- [rowAnnotation](#),
- [HeatmapAnnotation](#) and
- [Heatmap](#)

Usage

```
annotation_heatmap_exposures(
  in_exposures_df,
  in_annotation_df,
  in_annotation_col,
  in_signatures_ind_df,
  in_data_type = "norm exposures",
  in_method = "manhattan",
  in_palette = colorRamp2(c(0, 0.2, 0.4, 0.6), c("white", "yellow", "orange", "red")),
  in_cutoff = 0,
  in_filename = NULL,
  in_column_anno_borders = FALSE,
  in_row_anno_borders = FALSE,
  in_show_PIDs = TRUE,
  in_annotation_legend_side = "right"
)
```

Arguments

<code>in_exposures_df</code>	Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).
<code>in_annotation_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup for all layers of annotation
<code>in_annotation_col</code>	A list indicating colour attributions for all layers of annotation
<code>in_signatures_ind_df</code>	A data frame containing meta information about the signatures, especially the asserted colour
<code>in_data_type</code>	Title in the figure
<code>in_method</code>	Method of the clustering to be supplied to dist . Can be either of: euclidean, maximum, manhattan, canberra, binary or minkowski

in_palette	Palette with colours or colour codes for the heatmap. Default is <code>colorRamp2(c(0, 0.2, 0.4, 0.6), c('white', 'yellow', 'orange', 'red'))</code>
in_cutoff	A numeric value less than 1. Signatures from within <code>W</code> with an overall exposure less than <code>in_cutoff</code> will be discarded for the clustering.
in_filename	A path to save the heatmap. If none is specified, the figure will be plotted to the running environment.
in_column_anno_borders	Whether or not to draw separating lines between the fields in the annotation
in_row_anno_borders	Whether or not to draw separating lines between the fields in the annotation
in_show_PIDs	Whether or not to show the PIDs on the x-axis
in_annotation_legend_side	Where to put the legends of the annotation df, default is right.

Details

One additional parameter, `in_show_legend_bool_vector`, indicating which legends to display, is planned but deactivated in this version of the package. In order to use this features, it will be necessary to install the newest version of the packages **circlize** and **ComplexHeatmap** by Zuguang Gu: `devtools::install_github("jokergoo/circlize")` and `devtools::install_github("jokergoo/ComplexHeatmap")`

Value

The function doesn't return any value.

See Also

[Heatmap](#)
[complex_heatmap_exposures](#)

Examples

```
NULL
```

attribute_nucleotide_exchanges

Attribute the nucleotide exchange for an SNV

Description

SNVs are grouped into 6 different categories (12/2 as reverse complements are summed over). This function defines the attribution.

Usage

```
attribute_nucleotide_exchanges(  
  in_dat,  
  in_REF.field = "REF",  
  in_ALT.field = "ALT",  
  in_verbose = FALSE  
)
```

Arguments

<code>in_dat</code>	VRanges object or data frame which carries one column for the reference base and one column for the variant base
<code>in_REF.field</code>	String indicating which column of <code>in_dat</code> carries the reference base if dealing with data frames
<code>in_ALT.field</code>	String indicating which column of <code>in_dat</code> carries the variant base if dealing with data frames
<code>in_verbose</code>	Whether verbose or not.

Value

A character vector with as many rows as there are in `in_dat` which can be annotated (i.e. appended) to the input data frame.

Examples

```
test_df <- data.frame(
  CHROM=c(1,1,1,2,2,2,3,3,3,4,4,4,5,5),
  POS=c(1,2,3,4,5,6,1,2,3,4,5,6,7,8),
  REF=c("C","C","C","T","T","T","A","A","A","G","G","G","N","A"),
  ALT=c("A","G","T","A","C","G","C","G","T","A","C","T","A","N"))
test_df$change <- attribute_nucleotide_exchanges(
  test_df,in_REF.field = "REF",in_ALT.field = "ALT")
test_df
```

attribute_sequence_context_indel

Attribution of sequence context and size for an INDEL

Description

The function is a wrapper and uses [getSequenceContext](#) to annotate the sequence context.

Usage

```
attribute_sequence_context_indel(
  in_dat,
  in_REF.field = "REF",
  in_ALT.field = "ALT",
  in_verbose = FALSE,
  in_offsetL = 10,
  in_offsetR = 50
)
```

Arguments

<code>in_dat</code>	VRanges object or data frame which carries one column for the reference base and one column for the variant base
<code>in_REF.field</code>	String indicating which column of <code>in_dat</code> carries the reference base if dealing with data frames

<code>in_ALT.field</code>	String indicating which column of <code>in_dat</code> carries the variant base if dealing with data frames
<code>in_verbose</code>	Verbose if <code>in_verbose=1</code>
<code>in_offsetL</code>	Number of nucleotides which should be annotated downstream of the variant. Per default 10 bps are annotated
<code>in_offsetR</code>	Number of nucleotides which should be annotated upstream of the variant. Per default 50 bps are annotated

Value

VRanges object or data frame with the same number rows and additional columns containing the type of INDEL (Ins = insertion and Del = deletion), the annotated sequence context of the defined length, the absolute number of exchanged nucleotides and the nucleotide exchange between `in_REF.field` and `in_ALT.field`.

Examples

```
data(GenomeOfN1_raw)
GenomeOfN1_context <- attribute_sequence_context_indel(
  in_dat = head(GenomeOfN1_raw),
  in_REF.field = "REF",
  in_ALT.field = "ALT",
  in_verbose = FALSE,
  in_offsetL = 10, in_offsetR = 50)

GenomeOfN1_context
```

`attribution_of_indels` *Attribution of variant into one of the 83 INDEL categories*

Description

Each variant is categorized into one of the 83 INDEL categories. The classification likewise to Alexandrov et al., 2018 (<https://www.synapse.org/#!Synapse:syn11726616>). The number of 83 features are classified as follows:

1. Deletion of 1 bp C/(G) or T/(A) in a repetitive context. The context is classified into 1, 2, 3, 4, 5 or larger or equal to 6 times the same nucleotide(s).
2. Insertion of 1 bp C/(G) or T/(A) in a repetitive context. The context is classified into 0, 1, 2, 3, 4, or larger or equal to 5 times the same nucleotide(s).
3. Deletions of 2bps, 3bps, 4bps or more or equal to 5bps in a repetitive context. Each deletion is classified in a context of 1, 2, 3, 4, 5 or larger or equal to 6 times the same motif.
4. Insertion of 2 bps, 3 bps, 4 bps or more or equal to 5 bps in a repetitive context. Each deletion is classified in a context of 0, 1, 2, 3, 4 or larger or equal to 5 times the same motif.
5. Microhomology deletion of 2bps, 3bps, 4bps or more or equal to 5 bps in a partly repetitive context. The partly repetitive context is defined by motif length of minus 1 bp, 2 bps, 3 bps, 4 bps or more or equal to 5bps, which is located before and after the break-point junction of the deletion.

Usage

```
attribution_of_indels(in_dat_return = in_dat_return)
```

Arguments

`in_dat_return` Data frame constructed from a vcf-like file of a whole cohort or a single-sample. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or defined columns. One of these can carry a PID (patient or sample identifier) and subgroup information. Furthermore, the columns containing the sequence context and the absolute length of the INDEL as well as the INDEL type of the variant can be annotated to the vcf-like df with [attribute_sequence_context_indel](#). These columns are required to enable the construction of a mutational catalog.

Value

Data frame with the same dimension as the input data frame plus an additional column with the INDEL classification number corresponding to Alexandrov et al. 2018.

Examples

```
data(GenomeOfN1_raw)
GenomeOfN1_context <- attribute_sequence_context_indel(in_dat =
head(GenomeOfN1_raw))
GenomeOfN1_classified <- attribution_of_indels(GenomeOfN1_context)
GenomeOfN1_classified
```

```
build_gene_list_for_pathway
```

Build a gene list for a given pathway name

Description

Build a gene list for a given pathway name

Usage

```
build_gene_list_for_pathway(in_string, in_organism)
```

Arguments

`in_string` Name or description of the pathway
`in_organism` Name of the taxon to be searched in

Value

A character vector of gene names

See Also

[keggLink](#)
[keggFind](#)
[extract_names_from_gene_list](#)

Examples

```

species <- "hsa"
gene_lists_meta_df <- data.frame(
  name=c("BER", "NHEJ", "MMR"),
  explanation=c("base excision repair",
               "non homologous end joining",
               "mismatch repair"))
number_of_pathways <- dim(gene_lists_meta_df)[1]
gene_lists_list <- list()
for (i in seq_len(number_of_pathways)) {
  temp_list <-
    build_gene_list_for_pathway(gene_lists_meta_df$explanation[i],
                               species)
  gene_lists_list <- c(gene_lists_list, list(temp_list))
}
gene_lists_list

```

 classify_indels

INDEL function V1 - not compatible with AlexandrovSignatures

Description

INDEL function V1 - not compatible with AlexandrovSignatures

Usage

```

classify_indels(
  in_df,
  in_ALT.field = "ALT",
  in_REF.field = "REF",
  in_breaks = c(-Inf, -10, -3, 0, 2, 9, Inf),
  in_labels = c("del3", "del2", "del1", "in1", "in2", "in3")
)

```

Arguments

in_df	Input data frame containing the variances in a vcf-like format
in_ALT.field	Column number for alternative field
in_REF.field	Column number for reference field
in_breaks	Handed over to function cut
in_labels	Handed over to function cut

Value

classVector, a factor vector of indel sizes

Examples

NULL

compare_exposures	<i>Compares alternative exposures</i>
-------------------	---------------------------------------

Description

Compares exposures computed by two alternative approaches for the same cohort

Usage

```
compare_exposures(in_exposures1_df, in_exposures2_df, deselect_flag = TRUE)
```

Arguments

`in_exposures1_df` Numeric data frame with exposures, ideally the smaller exposure data is supplied first.

`in_exposures2_df` Numeric data frame with exposures, ideally the bigger exposure data is supplied second.

`deselect_flag` Whether signatures absent in both exposure data frames should be removed.

Value

A list with entries `merge_df`, `all_cor.coeff`, `all_p.value`, `cor.coeff_vector`, `p.value_vector`, `all_cor.test`, and `cor.test_list`.

- `merge_df`: Merged molten input exposure data frames
- `all_cor.coeff`: Pearson correlation coefficient for all data points, i.e. taken all signatures together
- `all_p.value`: P-value of the Pearson test for all data points, i.e. taken all signatures together
- `cor.coeff_vector`: A vector of Pearson correlation coefficients evaluated for every signature independently
- `p.value_vector`: A vector of p-values of the Pearson tests evaluated for every signature independently
- `all_cor.test`: A data structure as returned by `cor.test` for all data points, i.e. taken all signatures together
- `cor.test_list`: A list of data structures as returned by `cor.test`, but evaluated for every signature independently

Examples

NULL

compare_expousre_sets *Compare two sets of exposures by cosine distance*

Description

Compare two sets of exposures, stored in numerical data frames H1 and H2, by computing the row-wise cosine distance

Usage

```
compare_expousre_sets(in_df_small, in_df_big, in_distance = cosineDist)
```

Arguments

`in_df_small`, `in_df_big` Numerical data frames H1 and H2, ideally the bigger one first, both with `l` rows and `m1` and `m2` columns, `l` being the number of signatures and `m1` and `m2` being the respective numbers of samples or patient identifier of H1 and H2

`in_distance` A function which computes the distance measure, default is `cosineDist`

Value

A list with entries `distance`, `hierarchy_small` and `hierarchy_big`.

- `distance`: A numerical data frame with the cosine distances between the columns of H1, indexing the rows, and H2, indexing the columns
- `hierarchy_small`: A data frame carrying the information of ranked similarity between the signatures in H2 with the signatures in H1
- `hierarchy_big`: A data frame carrying the information of ranked similarity between the signatures in H1 with the signatures in H2

See Also

[cosineDist](#)

Examples

```
sig_1_df <- data.frame(matrix(c(1,0,0,0,0,1,0,0,0,0,1,0),ncol=3))
names(sig_1_df) <- paste0("B",seq_len(dim(sig_1_df)[2]))
sig_2_df <- data.frame(matrix(c(1,1,0,0,0,0,0,1,1),ncol=2))
compare_expousre_sets(sig_1_df,sig_2_df)
```

compare_sets	<i>Compare two sets of signatures by cosine distance</i>
--------------	--

Description

Compare two sets of signatures, stored in numerical data frames W1 and W2, by computing the column-wise cosine distance

Usage

```
compare_sets(in_df_small, in_df_big, in_distance = cosineDist)
```

Arguments

`in_df_small`, `in_df_big`
Numerical data frames W1 and W2, ideally the bigger one first, both with `n` rows and `l1` and `l2` columns, `n` being the number of features and `l1` and `l2` being the respective numbers of signatures of W1 and W2

`in_distance` A function which computes the distance measure, default is `cosineDist`

Value

A list with entries `distance`, `hierarchy_small` and `hierarchy_big`.

- `distance`: A numerical data frame with the cosine distances between the columns of W1, indexing the rows, and W2, indexing the columns
- `hierarchy_small`: A data frame carrying the information of ranked similarity between the signatures in W2 with the signatures in W1
- `hierarchy_big`: A data frame carrying the information of ranked similarity between the signatures in W1 with the signatures in W2

See Also

[cosineDist](#)

Examples

```
sig_1_df <- data.frame(matrix(c(1,0,0,0,0,1,0,0,0,0,1,0),ncol=3))
names(sig_1_df) <- paste0("B",seq_len(dim(sig_1_df)[2]))
sig_2_df <- data.frame(matrix(c(1,1,0,0,0,0,1,1),ncol=2))
compare_sets(sig_1_df,sig_2_df)
```

 compare_SMCs

Compare all strata from different stratifications

Description

Compare all strata from different orthogonal stratification axes, i.e. orthogonal SMCs by cosine similarity of signature exposures. First calls

- `make_strata_df`, then
- `plot_strata` and finally
- `make_comparison_matrix`

Usage

```
compare_SMCs(
  in_stratification_lists_list,
  in_signatures_ind_df,
  output_path,
  in_nrect = 5,
  in_attribute = ""
)
```

Arguments

<code>in_stratification_lists_list</code>	List of lists with entries from different (orthogonal) stratification axes or SMCs
<code>in_signatures_ind_df</code>	A data frame containing meta information about the signatures
<code>output_path</code>	Path to directory where the results, especially the figure produced by <code>corrplot</code> is going to be stored.
<code>in_nrect</code>	Number of clusters in the clustering procedure provided by <code>corrplot</code>
<code>in_attribute</code>	Additional string for the file name where the figure produced by <code>corrplot</code> is going to be stored.

Value

The comparison matrix of cosine similarities.

See Also

[plot_strata](#)
[make_comparison_matrix](#)

Examples

NULL

compare_to_catalogues *Compare one mutational catalogue to reference mutational catalogues*

Description

Compare one mutational catalogue (e.g. of one index patient) to a list of reference mutational catalogues (e.g. from the initial Alexandrov publication) by cosine similarities

Usage

```
compare_to_catalogues(in_index_df, in_comparison_list)
```

Arguments

`in_index_df` Data frame containing the mutational catalogue of interest

`in_comparison_list` List of data frames (ideally named) containing the reference mutational catalogues

Value

A similarity dataframe

Examples

```
NULL
```

complex_heatmap_exposures

Heatmap to cluster the PIDs on their signature exposures (Complex-Heatmap)

Description

The PIDs are clustered according to their signature exposures. uses package **ComplexHeatmap** by Zuguang Gu. This function calls:

- [rowAnnotation](#),
- [HeatmapAnnotation](#) and
- [Heatmap](#)

Usage

```

complex_heatmap_exposures(
  in_exposures_df,
  in_subgroups_df,
  in_signatures_ind_df,
  in_data_type = "norm exposures",
  in_method = "manhattan",
  in_subgroup_column = "subgroup",
  in_subgroup_colour_column = NULL,
  in_palette = colorRamp2(c(0, 0.2, 0.4, 0.6), c("white", "yellow", "orange", "red")),
  in_cutoff = 0,
  in_filename = NULL,
  in_column_anno_borders = FALSE,
  in_row_anno_borders = FALSE
)

```

Arguments

`in_exposures_df` Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).

`in_subgroups_df` A data frame indicating which PID (patient or sample identifier) belongs to which subgroup

`in_signatures_ind_df` A data frame containing meta information about the signatures, especially the asserted colour

`in_data_type` Title in the figure

`in_method` Method of the clustering to be supplied to `dist`. Can be either of: euclidean, maximum, manhattan, canberra, binary or minkowski

`in_subgroup_column` Indicates the name of the column in which the subgroup information is encoded in `in_subgroups_df`

`in_subgroup_colour_column` Indicates the name of the column in which the colour information for subgroups is encoded in `in_subgroups_df`. If NULL, a rainbow palette is used instead.

`in_palette` Palette with colours for the heatmap. Default is `colorRamp2(c(0, 0.2, 0.4, 0.6), c('white', 'yellow', 'orange', 'red'))`

`in_cutoff` A numeric value less than 1. Signatures from within W with an overall exposure less than `in_cutoff` will be discarded for the clustering.

`in_filename` A path to save the heatmap. If none is specified, the figure will be plotted to the running environment.

`in_column_anno_borders` Whether or not to draw separating lines between the fields in the annotation

`in_row_anno_borders` Whether or not to draw separating lines between the fields in the annotation

Details

It might be necessary to install the newest version of the development branch of the packages **cir-clize** and **ComplexHeatmap** by Zuguang Gu: `devtools::install_github("jokergoo/circlize")` and `devtools::install_github("jokergoo/ComplexHeatmap")`

Value

The function doesn't return any value.

See Also

[Heatmap](#)

Examples

```
data(lymphoma_cohort_LCD_results)
complex_heatmap_exposures(
  rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df,
  COSMIC_subgroups_df,
  chosen_signatures_indices_df,
  in_data_type="norm exposures",
  in_subgroup_colour_column="col",
  in_method="manhattan",
  in_subgroup_column="subgroup")
```

computeLogLik	<i>Compute the loglikelihood</i>
---------------	----------------------------------

Description

Compute the loglikelihood

Usage

```
computeLogLik(in_vector, in_pdf = NULL, verbose = FALSE)
```

Arguments

<code>in_vector</code>	Numeric vector of input values of which the loglikelihood is computed.
<code>in_pdf</code>	Probability distribution function, if NULL a normal distribution is used.
<code>verbose</code>	Verbose if <code>in_verbose=1</code>

Value

A numeric value (sum of the logarithms of the likelihoods of the input vector)

Examples

```
NULL
```

`compute_comparison_stat_df`*Extract statistical measures for entity comparison*

Description

Compare one mutational catalogue (e.g. of one index patient) to a list of reference mutational catalogues (e.g. from the initial Alexandrov publication) by cosine similarities

Usage`compute_comparison_stat_df(in_sim_df)`**Arguments**

`in_sim_df` A similarity data frame as extracted by [compare_to_catalogues](#)

Value

A dataframe containing statistical measures, prepared for bar plot

Examples`NULL`

`confidence_indel_calulation`*Wrapper to compute confidence intervals for SNV and INDEL signatures of a cohort or single-sample*

Description

Wrapper function around [confIntExp](#), which is applied to every signature or sample pair in a cohort. The extracted lower bound of the confidence intervals are added to the input data which is reodered and melted in order to prepare for visualization with ggplot2. The calculation of confidence intervals is based on a profiling likelihood algorithm and the wrapper calculates the data for the exposure contribution identified with SNV and INDEL signature decompositions and application of the following cutoffs:

1. CosmicValid_absCutoffVector
2. CosmicValid_normCutoffVector
3. CosmicArtif_absCutoffVector
4. CosmicArtif_normCutoffVector
5. PCAWGValidSNV_absCutoffVector
6. PCAWGValidID_absCutoffVector

The function makes use of different YAPSA functions. For each of the above stated cutoff vectors a per PID decomposition of the SNV and INDEL catalog is calculated respectively using `LCD_complex_cutoff_perPID`. In a next step, `variateExp` which is a wrapper around `confIntExp` to compute confidence intervals for a cohort is used. A dataframe is returned with the upper and lower bounds of the confidence intervals. In a last step `plotExposuresConfidence_indel` to plot the exposures to extracted signatures including confidence intervals computed with e.g. by `variateExp`.

Usage

```
confidence_indel_calulation(in_current_indel_df, in_current_snv_df)
```

Arguments

`in_current_indel_df`

A INDEL mutational catalog. Mutational catalog can be constructed with `create_indel_mutation_c`

`in_current_snv_df`

A SNV mutational catalog. Mutational catalog can be constructed with `create_mutation_catalogue`

Value

A list is returned containing 12 objects. For each cutoff data frame two corresponding object are present. First, the `p` table object which can be used for graphical visualization, and second a dataframe containing the corresponding upper and lower bounds of the confidence intervals.

Examples

```
data("GenomeOfN1_MutCat")
```

```
confidence_indel_only_calulation
```

Wrapper to compute confidence intervals for only INDEL signatures.

Description

Wrapper function around `confIntExp`, which is applied to every signature or sample pair in a cohort. The extracted lower bound of the confidence intervals are added to the input data which is re-ordered and melted in order to prepare for visualization with `ggplot2`. The calculation of confidence intervals is based on a profiling likelihood algorithm and the wrapper calculates the data for the exposure contribution identified with INDEL signature decomposition and the usage of `PCAWGValidID_absCutoffVector` data frame.

Usage

```
confidence_indel_only_calulation(in_current_indel_df)
```

Arguments

`in_current_indel_df`

A INDEL mutational catalog. Mutational catalog can be constructed with `create_indel_mutation_c`

Details

The function makes use of different YAPSA functions. For each of the above stated cutoff vectors a per PID decomposition of the SNV and INDEL catalog is calculated respectively using `LCD_complex_cutoff_perPID`. In a next step, `variateExp` which is a wrapper around `confIntExp` to compute confidence intervals for a cohort is used. A dataframe is returned with the upper and lower bounds of the confidence intervals. In a last step `plotExposuresConfidence_indel` to plot the exposures to extracted signatures including confidence intervals computed with e.g. by `variateExp`.

Value

A list is returned containing two objects. First, the plot object which can be used for graphical visualization, and second a dataframe containing the corresponding upper and lower bounds of the confidence intervals.

Examples

```
data("GenomeOfN1_MutCat")
temp_list <- confidence_indel_only_calculation(
  in_current_indel_df=MutCat_indel_df)
plot(temp_list$p_complete_PCAWG_ID)
head(temp_list$complete_PCAWG_ID)
```

confIntExp

Compute confidence intervals

Description

Compute confidence intervals using the (log-)likelihood ratio test, primarily for one input sample.

Usage

```
confIntExp(
  in_ind = 1,
  in_sigLevel = 0.05,
  in_delta = 1,
  in_exposure_vector = NULL,
  in_verbose = FALSE,
  ...
)
```

Arguments

<code>in_ind</code>	Index of the input signature to be variated.
<code>in_sigLevel</code>	Significance level (one-sided)
<code>in_delta</code>	Inflation parameter for the alternative model.
<code>in_exposure_vector</code>	Exposure vector computed for the input sample.
<code>in_verbose</code>	Whether to run verbose (TRUE) or not (FALSE)
<code>...</code>	Input parameters passed on to <code>variateExpSingle</code> .

Value

A list with entries

- upper: Upper bound of the confidence interval
- lower: Lower bound of the confidence interval

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
data(lymphoma_cohort_LCD_results)
data(sigs)
word_length <- 3
temp_list <- create_mutation_catalogue_from_df(
  lymphoma_test_df, this_seqnames.field = "CHROM",
  this_start.field = "POS", this_end.field = "POS",
  this_PID.field = "PID", this_subgroup.field = "SUBGROUP",
  this_refGenome = BSgenome.Hsapiens.UCSC.hg19,
  this_wordLength = word_length)
lymphoma_catalogue_df <- temp_list$matrix
lymphoma_PIDs <- colnames(lymphoma_catalogue_df)
data("lymphoma_cohort_LCD_results")
lymphoma_exposures_df <-
  lymphoma_Nature2013_COSMIC_cutoff_exposures_df[, lymphoma_PIDs]
lymphoma_sigs <- rownames(lymphoma_exposures_df)
lymphoma_sig_df <- AlexCosmicValid_sig_df[, lymphoma_sigs]
confIntExp(in_ind = 1, in_sigLevel = 0.05, in_delta = 0.4,
  in_exposure_vector = lymphoma_exposures_df[, 1],
  in_catalogue_vector = lymphoma_catalogue_df[, 1],
  in_sig_df = lymphoma_sig_df)
```

correct_rounded

Readjust the vector to it's original norm after rounding

Description

After use of the function [round_precision](#) the norm of the input vector may have been altered by the rounding procedure. This function restores the norm by altering only the largest entry in the rounded vector (in order to create the least possible relative error).

Usage

```
correct_rounded(x, in_interval = c(0, 1))
```

Arguments

x vector to be rounded
in_interval Interval

Value

The adapted form of the input vector x.

Examples

NULL

cosineDist *Compute the cosine distance of two vectors*

Description

Compute the cosine distance of two vectors

Usage

```
cosineDist(a, b)
```

Arguments

a, b Numerical vectors of same length

Value

The scalar product of the two input vectors divided by the product of the norms of the two input vectors

Examples

```
## 1. Orthogonal vectors:
cosineDist(c(1,0),c(0,1))
## 2. Non-orthogonal vectors:
cosineDist(c(1,0),c(1,1))
## Compare trigonometry:
1-cos(pi/4)
```

cosineMatchDist *Compute an altered cosine distance of two vectors*

Description

This is an altered cosine distance: it first reduced the dimension of the two input vectors to only those coordinates where both have non-zero entries. The cosine similarity is then computed on these reduced vectors, i.e. on a sub-vector space.

Usage

```
cosineMatchDist(a, b)
```

Arguments

a, b Numerical vectors of same length

Value

The scalar product of the reduced input vectors divided by the product of the norms of the two reduced input vectors

Examples

```
## 1. Orthogonal vectors:
cosineMatchDist(c(1,0),c(0,1))
## 2. Non-orthogonal vectors:
cosineMatchDist(c(1,0),c(1,1))
```

```
create_indel_mutation_catalogue_from_df
```

Wrapper to create an INDEL mutational catalog from a vcf-like data frame

Description

From data frame constructed from a vcf-file file the function `create_indel_mutation_catalogue_from_df` creates a mutational catalog `V` by sequentially applying the `attribute_sequence_context_indel`, `attribute_sequence_context_indel` and then `attribution_of_indels`. The runtime of the function is about 1 sec per 6 variants as sequence context as well as INDEL classification are timeconsuming to compute (optimization ongoing)

Usage

```
create_indel_mutation_catalogue_from_df(
  in_dat,
  in_signature_df,
  in_REF.field = "REF",
  in_ALT.field = "ALT",
  in_verbose = FALSE
)
```

Arguments

<code>in_dat</code>	A data frame constructed from a vcf-like file of a whole cohort or single-sample. The first columns are those of a standard vcf file (CHROM, POS, REF and ALT), followed by an arbitrary number of custom or used defined columns. One of these can carry a PID (patient or sample identifier) and one can carry subgroup information.
<code>in_signature_df</code>	A numeric data frame <code>W</code> with <code>n</code> rows and <code>l</code> columns, <code>n</code> being the number of features and <code>l</code> being the number of signatures. Data frame containing INDEL signatures which should be used to create the mutational catalogue <code>V</code> .
<code>in_REF.field</code>	String indicating which column of <code>in_dat</code> carries the reference base if dealing with data frames
<code>in_ALT.field</code>	String indicating which column of <code>in_dat</code> carries the variant base if dealing with data frames
<code>in_verbose</code>	Verbose if <code>in_verbose=1</code>

Value

A dataframe in the format of a mutational catalog V , which can be used for [LCD](#) analysis

Examples

```
data(sigs_pcawg)
data(GenomeOfNI_raw)
temp_df <- translate_to_hg19(GenomeOfNI_raw[1:200,], "CHROM")
temp_df$PID <- sample(c("PID1", "PID2", "PID3", "PID4", "PID5"), 200, replace=TRUE)
temp <- create_indel_mutation_catalogue_from_df(in_dat = temp_df,
  in_signature_df = PCAWG_SP_ID_sigs_df,
  in_REF.field = "REF",
  in_ALT.field = "ALT",
  in_verbose = FALSE)
dim(temp)
head(temp)
```

```
create_indel_mut_cat_from_df
```

Create a Mutational catalog from a data frame

Description

This function creates a mutational catalog from a data frame. It requires the returned data frame obtained with [attribution_of_indels](#).

Usage

```
create_indel_mut_cat_from_df(in_df, in_signatures_df)
```

Arguments

in_df A data frame constructed from a vcf-like file of a whole cohort or single-sample. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or user-defined columns. One of these can carry a PID (patient or sample identifier) and the subgroup information. Additionally to construct the mutational catalog each variant needs to be characterized into one of the 83 INDEL feature classes, which can be performed with [attribution_of_indels](#)

in_signatures_df

A numeric data frame W with n rows and l columns, n being the number of features and l being the number of signatures. Data frame containing INDEL signatures which should be used to create the mutational catalog V .

Value

A count dataframe, the mutational catalog V with rownames indicating the INDELS and colnames having the PIDs

Examples

```

data(GenomeOfNI_raw)
data(sigs_pcawg)
GenomeOfNI_context <- attribute_sequence_context_indel(in_dat =
  head(GenomeOfNI_raw))
GenomeOfNI_classified <- attribution_of_indels(GenomeOfNI_context)
GenomeOfNI_mut_cat <- create_indel_mut_cat_from_df(GenomeOfNI_classified,
  in_signatures_df=PCAWG_SP_ID_sigs_df)

```

```
create_mutation_catalogue_from_df
```

Create a Mutational Catalogue from a data frame

Description

This function creates a mutational catalogue from a data frame. It is a wrapper function for [create_mutation_catalogue_from_VR](#): it first creates a `VRanges` object from the data frame by [makeVRangesFromDataFrame](#) and then passes this object on to the above mentioned custom function.

Usage

```

create_mutation_catalogue_from_df(
  this_df,
  this_refGenome_Seqinfo = NULL,
  this_seqnames.field = "X.CHROM",
  this_start.field = "POS",
  this_end.field = "POS",
  this_PID.field = "PID",
  this_subgroup.field = "subgroup",
  this_refGenome,
  this_wordLength,
  this_verbose = 1,
  this_rownames = c(),
  this_adapt_rownames = 1
)

```

Arguments

<code>this_df</code>	A data frame constructed from a vcf-like file of a whole cohort. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or used defined columns. One of these can carry a PID (patient or sample identifier) and one can carry subgroup information.
<code>this_refGenome_Seqinfo</code>	A <code>seqInfo</code> object, referring to the reference genome used. Argument passed on to makeGRangesFromDataFrame and thus indirectly to makeGRangesFromDataFrame .
<code>this_seqnames.field</code>	Indicates the name of the column in which the chromosome is encoded
<code>this_start.field</code>	Indicates the name of the column in which the start coordinate is encoded

<code>this_end.field</code>	Indicates the name of the column in which the end coordinate is encoded
<code>this_PID.field</code>	Indicates the name of the column in which the PID (patient or sample identifier) is encoded
<code>this_subgroup.field</code>	Indicates the name of the column in which the subgroup information is encoded
<code>this_refGenome</code>	The reference genome handed over to <code>create_mutation_catalogue_from_VR</code> and indirectly to <code>mutationContext</code> and used to extract the motif context of the variants in <code>in_vr</code> .
<code>this_wordLength</code>	The size of the motifs to be extracted by <code>mutationContext</code>
<code>this_verbose</code>	Verbose if <code>this_verbose=1</code>
<code>this_rownames</code>	Optional parameter to specify rownames of the mutational catalogue V i.e. the names of the features.
<code>this_adapt_rownames</code>	Rownames of the output matrix will be adapted if <code>this_adapt_rownames=1</code>

Value

A list with entries `matrix` and `frame` obtained from `create_mutation_catalogue_from_VR`:

- `matrix`: The mutational catalogue V
- `frame`: Additional and meta information on rownames (features), colnames (PIDs) and subgroup attribution.

See Also

[makeVRangesFromDataFrame](#)

[create_mutation_catalogue_from_VR](#)

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
word_length <- 3
temp_list <- create_mutation_catalogue_from_df(
  lymphoma_test_df, this_seqnames.field = "CHROM",
  this_start.field = "POS", this_end.field = "POS",
  this_PID.field = "PID", this_subgroup.field = "SUBGROUP",
  this_refGenome = BSgenome.Hsapiens.UCSC.hg19,
  this_wordLength = word_length)
dim(temp_list$matrix)
head(temp_list$matrix)
```

```
create_mutation_catalogue_from_VR
```

Create a Mutational Catalogue from a VRanges Object

Description

This function creates a mutational catalogue from a VRanges Object by first calling [mutationContext](#) to establish the motif context of the variants in the input VRanges and then calling [motifMatrix](#) to build the mutational catalogue V.

Usage

```
create_mutation_catalogue_from_VR(
  in_vr,
  in_refGenome,
  in_wordLength,
  in_PID.field = "PID",
  in_verbose = 0,
  in_rownames = c(),
  adapt_rownames = 1
)
```

Arguments

<code>in_vr</code>	A VRanges object constructed from a vcf-like file of a whole cohort. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or used defined columns. One of these can carry a PID (patient or sample identifier) and one can carry subgroup information.
<code>in_refGenome</code>	The reference genome handed over to mutationContext and used to extract the motif context of the variants in <code>in_vr</code> .
<code>in_wordLength</code>	The size of the motifs to be extracted by mutationContext
<code>in_PID.field</code>	Indicates the name of the column in which the PID (patient or sample identifier) is encoded
<code>in_verbose</code>	Verbose if <code>in_verbose=1</code>
<code>in_rownames</code>	Optional parameter to specify rownames of the mutational catalogue V i.e. the names of the features.
<code>adapt_rownames</code>	Rownames of the output matrix will be adapted if <code>adapt_rownames=1</code>

Value

A list with entries `matrix`, `frame`,

- `matrix`: The mutational catalogue V
- `frame`: Additional and meta information on rownames (features), colnames (PIDs) and subgroup attribution.

See Also

[mutationContext](#)

[motifMatrix](#)

Examples

```

library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
data(sigs)
word_length <- 3
temp_vr <- makeVRangesFromDataFrame(
  lymphoma_test_df, in_seqnames.field="CHROM",
  in_subgroup.field="SUBGROUP", verbose_flag=1)
temp_list <- create_mutation_catalogue_from_VR(
  temp_vr, in_refGenome=BSgenome.Hsapiens.UCSC.hg19,
  in_wordLength=word_length, in_PID.field="PID",
  in_verbose=1)
dim(temp_list$matrix)
head(temp_list$matrix)
test_list <- split(lymphoma_test_df, f=lymphoma_test_df$PID)
other_list <- list()
for(i in seq_len(length(test_list))){
  other_list[[i]] <- test_list[[i]][c(1:80),]
}
other_df <- do.call(rbind, other_list)
other_vr <- makeVRangesFromDataFrame(
  other_df, in_seqnames.field="CHROM",
  in_subgroup.field="SUBGROUP", verbose_flag=1)
other_list <- create_mutation_catalogue_from_VR(
  other_vr, in_refGenome=BSgenome.Hsapiens.UCSC.hg19,
  in_wordLength=word_length, in_PID.field="PID",
  in_verbose=1, in_rownames=rownames(AlexCosmicValid_sig_df))
dim(other_list$matrix)
head(other_list$matrix)

```

cutoffs

Cutoffs for a supervised analysis of mutational signatures.

Description

Series of data frames with signature-specific cutoffs. All values represent optimal cutoffs. The optimal cutoffs were determined for different choices of parameters in the cost function of the optimization. The row index is equivalent to the ratio between costs for false negative attribution and false positive attribution. The columns correspond to the different signatures. To be used with [LCD_complex_cutoff](#). There are two different sets of cutoffs one for the signatures described by Alexandrov et al. (Natue 2013) and one for the signatures documented in Alexandriv et al. (biorxiv 2018). The calculation of the PCAWG signature specific cutoffs was performed in a single-sample resolution which are both valid for whole genome and whole exome sequencing data analysis.

`cutoffCosmicValid_rel_df`: Optimal cutoffs for [AlexCosmicValid_sig_df](#), i.e. COSMIC signatures, only validated, trained on relative exposures.

`cutoffCosmicArtif_rel_df`: Optimal cutoffs for [AlexCosmicArtif_sig_df](#), i.e. COSMIC signatures, including artifact signatures, trained on relative exposures.

`cutoffCosmicValid_abs_df`: Optimal cutoffs for [AlexCosmicValid_sig_df](#), i.e. COSMIC signatures, only validated, trained on absolute exposures.

`cutoffCosmicArtif_abs_df`: Optimal cutoffs for [AlexCosmicArtif_sig_df](#), i.e. COSMIC signatures, including artifact signatures, trained on absolute exposures.

cutoffInitialValid_rel_df: Optimal cutoffs for [AlexInitialValid_sig_df](#), i.e. initially published signatures, only validated signatures, trained on relative exposures.

cutoffInitialArtif_rel_df: Optimal cutoffs for [AlexInitialArtif_sig_df](#), i.e. initially published signatures, including artifact signatures, trained on relative exposures.

cutoffInitialValid_abs_df: Optimal cutoffs for [AlexInitialValid_sig_df](#), i.e. initially published signatures, only validated signatures, trained on absolute exposures.

cutoffInitialArtif_abs_df: Optimal cutoffs for [AlexInitialArtif_sig_df](#), i.e. initially published signatures, including artifact signatures, trained on absolute exposures.

Usage

```
data(cutoffs)
```

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

cutoffs_pcawg

Opt. cutoffs, PCAWG SNV signatures, including artifacts

Description

cutoffPCAWG_SBS_WGSWES_artifPid_df: Optimal cutoffs for [PCAWG_SP_SBS_sigs_Artif_df](#), i.e. initially published signatures, including artifact signatures, trained in a single-sample resolution.

cutoffPCAWG_SBS_WGSWES_realPid_df: Optimal cutoffs for [PCAWG_SP_SBS_sigs_Real_df](#), i.e. initially published signatures, only validated signatures, trained in a single-sample resolution.

cutoffPCAWG_ID_WGS_Pid_df: Optimal cutoffs for [PCAWG_SP_ID_sigs_df](#), i.e. initially published signatures, signatures, trained in a single-sample resolution.

Usage

```
data(cutoffs_pcawg)
```

Author(s)

Lea Jopp-Saile <huebschmann.daniel@gmail.com>

`cut_breaks_as_intervals`*Wrapper for cut*

Description

In this wrapper function for the known `cut` function, the breaks vector need not be supplied directly, instead, for every break, an interval is supplied and the function optimizes the choice of the breakpoint by choosing a local minimum of the distribution.

Usage

```
cut_breaks_as_intervals(  
  in_vector,  
  in_outlier_cutoffs = c(0, 3000),  
  in_cutoff_ranges_list = list(c(60, 69), c(25, 32)),  
  in_labels = c("late", "intermediate", "early"),  
  in_name = "",  
  output_path = NULL  
)
```

Arguments

<code>in_vector</code>	Vector of numerical continuously distributed input
<code>in_outlier_cutoffs</code>	Interval specifying the upper and lower bounds of the range to be considered
<code>in_cutoff_ranges_list</code>	List of intervals in which the cutoffs for <code>cut</code> have to be optimized.
<code>in_labels</code>	Labels assigned to the strata or factors returned
<code>in_name</code>	String specifying the name of the quantity analyzed (and plotted on the x-axis of the figure to be created).
<code>output_path</code>	Path where the figure produced by the density function should be stored if non-NULL.

Value

A list with entries `category_vector`, `density_plot` and `cutoffs`

- `category_vector`: Factor vector of the categories or strata, of the same length as `in_vector`
- `density_plot`: Density plot produced by the density function and indication of the chosen cutoffs.
- `cutoffs`: Vector of the computed optimal cutoffs

See Also

[cut](#)

[density](#)

Examples

```

data(lymphoma_test)
lymphoma_test_df$random_norm <- rnorm(dim(lymphoma_test_df)[1])
temp_list <- cut_breaks_as_intervals(
  lymphoma_test_df$random_norm,
  in_outlier_cutoffs=c(-4,4),
  in_cutoff_ranges_list=list(c(-2.5,-1.5),c(0.5,1.5)),
  in_labels=c("small","intermediate","big"))
temp_list$density_plot

```

deriveSigInd_df	<i>Derive a signature_indices_df object</i>
-----------------	---

Description

Derive a data frame of type signature_indices_df (additional information for a set of signatures) from a set of given signatures for a set of new signatures.

Usage

```
deriveSigInd_df(querySigs, subjectSigs, querySigInd = NULL, in_sort = FALSE)
```

Arguments

querySigs	The signatures to compare to (given signatures).
subjectSigs	The signatures to be compared (new signatures). Alternatively this may be a complex object of type list and contain data from different deconvolutions, each of which having a set of signatures to be compared.
querySigInd	The object of type signature_indices_df (additional informatio for a set of signatures) belonging to the set of known signatures.
in_sort	Whether to sort or not

Value

An object of type signature_indices_df (additional informatio for a set of signatures) belonging to the set of new signatures.

See Also

[relateSigs](#)

Examples

```
NULL
```

disambiguateVector *Disambiguate a vector*

Description

Add numbered suffixes to redundant entries in a vector

Usage

```
disambiguateVector(in_vector)
```

Arguments

in_vector Input vector

Value

The disambiguated vector.

Examples

```
NULL
```

enrichSigs *Compare to background distribution*

Description

Compare exposures from an analysis of mutational signatures in a cohort of interest to exposures computed in a background (e.g. the set of WES and WGS samples from Alexandrov 2013).

Usage

```
enrichSigs(in_cohort_exposures_df, in_background_exposures_df, in_sig_df)
```

Arguments

in_cohort_exposures_df Numerical data frame of the exposures of the cohort of interest.
in_background_exposures_df Numerical data frame of the exposures of the background.
in_sig_df Numerical data frame encoding the mutational signatures.

Value

A data frame with counts and p-values from Fisher tests.

Examples

```
NULL
```

exampleINDEL_YAPSA	<i>Data structures used in examples, Indel tests and the Indel signature vignette of the YAPSA package.</i>
--------------------	---

Description

Data structures used in examples, Indel tests and the Indel signature vignette of the YAPSA package.

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

References

<https://www.ncbi.nlm.nih.gov/pubmed/23945592>

exampleYAPSA	<i>Test and example data</i>
--------------	------------------------------

Description

Data structures used in examples, SNV tests and the SNV signature vignette of the YAPSA package.

lymphoma_PID_df: A data frame carrying subgroup information for a subcohort of samples used in the vignette. Data in the vignette is downloaded from ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt. In the file available under that link somatic point mutation calls from several samples are listed in a vcf-like format. One column encodes the sample the variant was found in. In the vignette we want to restrict the analysis to only a fraction of these involved samples. The data frame lymphoma_PID_df carries the sample identifiers (PID) as rownames and the attributed subgroup in a column called subgroup.

lymphoma_test_df: A data frame carrying point mutation calls. It represents a subset of the data stored in ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt. In the file available under that link somatic point mutation calls from several samples are listed in a vcf-like format. One column encodes the sample the variant was found in. The data frame lymphoma_test_df has only the variants occurring in the sample identifiers (PIDs) 4112512, 4194218 and 4121361.

lymphoma_Nature2013_raw_df: A data frame carrying point mutation calls. It represents a subset of the data stored in ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt. In the file available under that link somatic point mutation calls from several samples are listed in a vcf-like format. One column encodes the sample the variant was found in.

lymphoma_Nature2013_COSMIC_cutoff_exposures_df: Data frame with exposures for testing the plot functions. Data taken from ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt.

rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df: Data frame with normalized or relative exposures for testing the plot functions. Data taken from <ftp://ftp.sanger.ac.uk/pub/>

[cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt](#).

COSMIC_subgroups_df: Subgroup information for the data stored in [lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#) and [rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#).

chosen_AlexInitialArtif_sigInd_df: Signature information for the data stored in [lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#) and [rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#).

chosen_signatures_indices_df: Signature information for the data stored in [lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#) and [rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df](#).

Usage

```
data(lymphoma_PID)

data(lymphoma_test)

data(lymphoma_Nature2013_raw)

data(lymphoma_cohort_LCD_results)

data(lymphoma_cohort_LCD_results)

data(lymphoma_cohort_LCD_results)

data(lymphoma_cohort_LCD_results)

data(lymphoma_cohort_LCD_results)
```

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

References

<https://www.ncbi.nlm.nih.gov/pubmed/23945592>

Examples

```
data(lymphoma_test)
head(lymphoma_test_df)
dim(lymphoma_test_df)
table(lymphoma_test_df$PID)

data(lymphoma_Nature2013_raw)
head(lymphoma_Nature2013_raw_df)
dim(lymphoma_Nature2013_raw_df)
```

`exchange_colour_vector`*Colours codes for displaying SNVs*

Description

Vector attributing colours to nucleotide exchanges used when displaying SNV information, e.g. in a rainfall plot.

Usage

```
data(exchange_colour_vector)
```

Value

A named character vector

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

`exome_mutCatRaw_df`*Example mutational catalog for the exome vignette*

Description

`exome_mutCatRaw_df`: A data frame in the format of a SNV mutation catalog. The mutational catalog contains SNV variants from a cohort of small-cell lung cancer published by Rudin et al. (Nature Genetics 2012) which was later used in the de novo discovery analysis of mutational signatures in human cancer by Alexandrov et al. (Nature 2013).

Usage

```
data(smallCellLungCancerMutCat_NatureGenetics2012)
```

Value

A data frame in the layout of a SNV mutational catalog

References

<https://www.nature.com/articles/ng.2405>

Examples

```
data(smallCellLungCancerMutCat_NatureGenetics2012)
head(exome_mutCatRaw_df)
dim(exome_mutCatRaw_df)
```

exposures_barplot *Wrapper for enhanced_barplot*

Description

Wrapper for enhanced_barplot

Usage

```
exposures_barplot(
  in_exposures_df,
  in_signatures_ind_df = NULL,
  in_subgroups_df = NULL,
  in_sum_ind = NULL,
  in_subgroups.field = "subgroup",
  in_title = "",
  in_labels = TRUE,
  in_show_subgroups = TRUE,
  ylab = NULL,
  in_barplot_borders = TRUE,
  in_column_anno_borders = FALSE
)
```

Arguments

`in_exposures_df` Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).

`in_signatures_ind_df` A data frame containing meta information about the signatures. If NULL, the colour information for the signatures is taken from a rainbow palette.

`in_subgroups_df` A data frame indicating which PID (patient or sample identifier) belongs to which subgroup. If NULL, it is assumed that all PIDs belong to one common subgroup. The colour coding for the default subgroup is red.

`in_sum_ind` Index vector influencing the order in which the PIDs are going to be displayed

`in_subgroups.field` String indicating the column name in `in_subgroups_df` to take the subgroup information from.

`in_title` Title for the plot to be created.

`in_labels` Flag, if TRUE the PIDs are displayed on the x-axis

`in_show_subgroups` Flag, if TRUE then PIDs are grouped by subgroups

`ylab` Label of the y-axis on the plot to be generate

`in_barplot_borders` Whether or not to show border lines in barplot

`in_column_anno_borders` Whether or not to draw separating lines between the fields in the annotation

Value

The generated barplot - a ggplot2 plot

Examples

```
data(lymphoma_cohort_LCD_results)
exposures_barplot(lymphoma_Nature2013_COSMIC_cutoff_exposures_df,
                  chosen_signatures_indices_df,
                  COSMIC_subgroups_df)
```

extract_names_from_gene_list

Return gene names from gene lists

Description

Return gene names from gene lists

Usage

```
extract_names_from_gene_list(in_KEGG_gene_list, 1)
```

Arguments

in_KEGG_gene_list	Gene list to extract names from
1	Index of the gene to be extracted

Value

The gene name.

See Also

[keggGet](#)

[build_gene_list_for_pathway](#)

Examples

```
NULL
```

`find_affected_PIDs` *Find samples affected*

Description

Find samples affected by SNVs in a certain pathway

Usage

```
find_affected_PIDs(in_gene_list, in_gene_vector, in_PID_vector)
```

Arguments

`in_gene_list` List of genes in the pathway of interest.
`in_gene_vector` Character vector for genes annotated to SNVs as in `vcf_like_df`.
`in_PID_vector` Character vector for sample names annotated to SNVs as in `vcf_like_df`.

Value

A character vector of the names of the affected samples

Examples

```
NULL
```

`GenomeOfNL_raw` *Example data for the Indel vignette*

Description

`GenomeOfNL_raw`: A data frame contains the gemiline varinats of the dutch population. carrying point mutation calls. It represents a subset of the data stored in ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt. In the file available under that link somatic point mutation calls from several samples are listed in a vcf-like format. One column encodes the sample the variant was found in.

Usage

```
data(GenomeOfNL_raw)
```

Value

A data frame in a vcf-like format

References

release version 5 https://www.nlgenome.nl/menu/main/app-go-nl/?page_id=9

Examples

```
data(GenomeOfNI_raw)
head(GenomeOfNI_raw)
dim(GenomeOfNI_raw)
```

getSequenceContext	<i>Extracts the sequence context up and downstream of a nucleotide position</i>
--------------------	---

Description

Extracts the sequence context up and downstream of a nucleotide position

Usage

```
getSequenceContext(position, chr, offsetL = 10, offsetR = 50)
```

Arguments

position	Start position of the considered INDEL
chr	Chromosome of the considered INDEL
offsetL	Number of nucleotides downstream of position
offsetR	Number of nucleotides upstream of position

Value

Returns a character string containing the defined sequence context

Examples

```
library(Biostrings)
library(BSgenome.Hsapiens.UCSC.hg19)

sequence_context <- getSequenceContext(position = 123456789, chr = "chr12",
                                       offsetL= 10, offsetR=50)

sequence_context
```

get_extreme_PIDs	<i>Return those PIDs which have an extreme pattern for signature exposure</i>
------------------	---

Description

For all signatures found in a project, this function returns the sample identifiers (PIDs) with extremely high or extremely low exposures of the respective signatures.

Usage

```
get_extreme_PIDs(in_exposures_df, in_quantile = 0.03)
```

Arguments

in_exposures_df	Data frame with the signature exposures
in_quantile	Quantile for the amount of extreme PIDs to be selected.

Value

A data frame with 4 rows per signature (high PIDs, high exposures, low PIDs, low exposures); the number of columns depends on the quantile chosen.

Examples

```
data(lymphoma_cohort_LCD_results)
get_extreme_PIDs(lymphoma_Nature2013_COSMIC_cutoff_exposures_df, 0.05)
```

hclust_exposures	<i>Cluster the PIDs according to their signature exposures</i>
------------------	--

Description

The PIDs are clustered according to their signature exposures by calling first creating a distance matrix:

- `dist`, then
- `hclust` and then
- `labels_colors` to colour the labels (the text) of the leaves in the dendrogram.

Typically one colour per subgroup.

Usage

```

hclust_exposures(
  in_exposures_df,
  in_subgroups_df,
  in_method = "manhattan",
  in_subgroup_column = "subgroup",
  in_palette = NULL,
  in_cutoff = 0,
  in_filename = NULL,
  in_shift_factor = 0.3,
  in_cex = 0.2,
  in_title = "",
  in_plot_flag = FALSE
)

```

Arguments

<code>in_exposures_df</code>	Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).
<code>in_subgroups_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup
<code>in_method</code>	Method of the clustering to be supplied to <code>dist</code> . Can be either of: euclidean, maximum, manhattan, canberra, binary or minkowski
<code>in_subgroup_column</code>	Indicates the name of the column in which the subgroup information is encoded in <code>in_subgroups_df</code>
<code>in_palette</code>	Palette with colours or colour codes for the labels (the text) of the leaves in the dendrogram. Typically one colour per subgroup. If none is specified, a rainbow palette of the length of the number of subgroups will be used as default.
<code>in_cutoff</code>	A numeric value less than 1. Signatures from within W with an overall exposure less than <code>in_cutoff</code> will be discarded for the clustering.
<code>in_filename</code>	A path to save the dendrogram. If none is specified, the figure will be plotted to the running environment.
<code>in_shift_factor</code>	Graphical parameter to adjust figure to be created
<code>in_cex</code>	Graphical parameter to adjust figure to be created
<code>in_title</code>	Title in the figure to be created under <code>in_filename</code>
<code>in_plot_flag</code>	Whether or not to display the dendrogram

Value

A list with entries `hclust` and `dendrogram`.

- `hclust`: The object created by `hclust`
- `dendrogram`: The above object wrapped in `as.dendrogram`

See Also

[hclust](#)
[dist](#)
[labels_colors](#)

Examples

```
data(lymphoma_cohort_LCD_results)
hclust_exposures(rel_lymphoma_Nature2013_COSMIC_cutoff_exposures_df,
                 COSMIC_subgroups_df,
                 in_method="manhattan",
                 in_subgroup_column="subgroup")
```

 LCD

Linear Combination Decomposition

Description

LCD performs a mutational signatures decomposition of a given mutational catalogue V with known signatures W by solving the minimization problem $\min(\|W * H - V\|)$ with additional constraints of non-negativity on H where W and V are known

Usage

```
LCD(in_mutation_catalogue_df, in_signatures_df, in_per_sample_cutoff = 0)
```

Arguments

`in_mutation_catalogue_df`
 A numeric data frame V with n rows and m columns, n being the number of features and m being the number of samples

`in_signatures_df`
 A numeric data frame W with n rows and l columns, n being the number of features and l being the number of signatures

`in_per_sample_cutoff`
 A numeric value less than 1. Signatures from within W with an exposure per sample less than `in_cutoff` will be discarded.

Value

The exposures H , a numeric data frame with l rows and m columns, l being the number of signatures and m being the number of samples

See Also

[lsei](#)

Examples

```

## define raw data
W_prim <- matrix(c(1,2,3,4,5,6),ncol=2)
W_prim_df <- as.data.frame(W_prim)
W_df <- YAPSA::normalize_df_per_dim(W_prim_df,2) # corresponds to the sigs
W <- as.matrix(W_df)
## 1. Simple case: non-negativity already in raw data
H <- matrix(c(2,5,3,6,1,9,1,2),ncol=4)
H_df <- as.data.frame(H) # corresponds to the exposures
V <- W %*% H # matrix multiplication
V_df <- as.data.frame(V) # corresponds to the mutational catalogue
exposures_df <- YAPSA::LCD(V_df,W_df)
## 2. more complicated: raw data already contains negative elements
## define indices where sign is going to be swapped
sign_ind <- c(5,7)
## now compute the indices of the other fields in the columns affected
## by the sign change
row_ind <- sign_ind % dim(H)[1]
temp_ind <- 2*row_ind -1
other_ind <- sign_ind + temp_ind
## alter the matrix H to yield a new mutational catalogue
H_compl <- H
H_compl[sign_ind] <- (-1)*H[sign_ind]
H_compl_df <- as.data.frame(H_compl) # corresponds to the exposures
V_compl <- W %*% H_compl # matrix multiplication
V_compl_df <- as.data.frame(V_compl) # corresponds to the mutational catalog
exposures_df <- YAPSA::LCD(V_compl_df,W_df)
exposures <- as.matrix(exposures_df)

```

LCD_complex_cutoff *LCD with a signature-specific cutoff on exposures*

Description

LCD_cutoff performs a mutational signatures decomposition by Linear Combination Decomposition (LCD) of a given mutational catalogue V with known signatures W by solving the minimization problem $\min(\|W * H - V\|)$ with additional constraints of non-negativity on H where W and V are known, but excludes signatures with an overall contribution less than a given signature-specific cutoff (and thereby accounting for a background model) over the whole cohort.

[LCD_complex_cutoff_perPID](#) is a wrapper for [LCD_complex_cutoff](#) and runs individually for every PID.

[LCD_extractCohort_callPerPID](#) runs [LCD_complex_cutoff](#) and takes the identified signatures as input for [LCD_complex_cutoff_perPID](#).

[LCD_complex_cutoff_consensus](#) calls [LCD_complex_cutoff_combined](#) AND [LCD_complex_cutoff_perPID](#) and makes a consensus signature call set.

[LCD_complex_cutoff_combined](#) is a wrapper for [LCD_complex_cutoff](#), [LCD_complex_cutoff_perPID](#), [LCD_complex_cutoff_consensus](#) AND [LCD_extractCohort_callPerPID](#).

Usage

```
LCD_complex_cutoff(  
  in_mutation_catalogue_df,  
  in_signatures_df,  
  in_cutoff_vector = NULL,  
  in_filename = NULL,  
  in_method = "abs",  
  in_per_sample_cutoff = 0,  
  in_rescale = TRUE,  
  in_sig_ind_df = NULL,  
  in_cat_list = NULL  
)  
  
LCD_complex_cutoff_perPID(  
  in_mutation_catalogue_df,  
  in_signatures_df,  
  in_cutoff_vector = NULL,  
  in_filename = NULL,  
  in_method = "abs",  
  in_rescale = TRUE,  
  in_sig_ind_df = NULL,  
  in_cat_list = NULL,  
  minimumNumberOfAlterations = 25  
)  
  
LCD_extractCohort_callPerPID(  
  in_mutation_catalogue_df,  
  in_signatures_df,  
  in_cutoff_vector = NULL,  
  in_filename = NULL,  
  in_method = "abs",  
  in_rescale = TRUE,  
  in_sig_ind_df = NULL,  
  in_cat_list = NULL,  
  in_verbose = FALSE,  
  minimumNumberOfAlterations = 25,  
  cutoff_type = "adaptive"  
)  
  
LCD_complex_cutoff_consensus(  
  in_mutation_catalogue_df = NULL,  
  in_signatures_df = NULL,  
  in_cutoff_vector = NULL,  
  in_filename = NULL,  
  in_method = "abs",  
  in_rescale = TRUE,  
  in_sig_ind_df = NULL,  
  in_cat_list = NULL,  
  in_cohort_LCDlist = NULL,  
  in_perPID_LCDlist = NULL,  
  addSigs_cohort_cutoff = 0.25,  
  addSigs_perPID_cutoff = 0.25,
```

```

    addSigs_relAbs_cutoff = 0.01,
    keep.unassigned = FALSE,
    keep.all.cohort.sigs = TRUE,
    in_verbose = FALSE,
    minimumNumberOfAlterations = 25
)

LCD_complex_cutoff_combined(
  in_mutation_catalogue_df = NULL,
  in_signatures_df = NULL,
  in_cutoff_vector = NULL,
  in_filename = NULL,
  in_method = "abs",
  in_rescale = TRUE,
  in_sig_ind_df = NULL,
  in_cat_list = NULL,
  addSigs_cohort_cutoff = 0.25,
  addSigs_perPID_cutoff = 0.25,
  addSigs_relAbs_cutoff = 0.01,
  keep.all.cohort.sigs = TRUE,
  in_verbose = FALSE,
  minimumNumberOfAlterations = 25,
  cutoff_type = "adaptive"
)

```

Arguments

in_mutation_catalogue_df
A numeric data frame V with n rows and m columns, n being the number of features and m being the number of samples

in_signatures_df
A numeric data frame W with n rows and l columns, n being the number of features and l being the number of signatures

in_cutoff_vector
A numeric vector of values less than 1. Signatures from within W with an overall exposure less than the respective value in in_cutoff_vector will be discarded.

in_filename
A path to generate a histogram of the signature exposures if non-NULL

in_method
Indicate to which data the cutoff shall be applied: absolute exposures, relative exposures

in_per_sample_cutoff
A numeric value less than 1. Signatures from within W with an exposure per sample less than in_cutoff will be discarded.

in_rescale
Boolean, if TRUE (default) the exposures are rescaled such that colSums over exposures match colSums over mutational catalogue

in_sig_ind_df
Data frame of type signature_indices_df, i.e. indicating name, function and meta-information of the signatures. Default is NULL.

in_cat_list
List of categories for aggregation. Have to be among the column names of in_sig_ind_df. Default is NULL.

minimumNumberOfAlterations
The perPID part of the analysis issues a warning if one sample has less mutations than this minimum cutoff.

<code>in_verbose</code>	Verbose if <code>in_verbose=1</code>
<code>cutoff_type</code>	If chosen to be "adaptive", the default, then signature-specific cutoffs are used for the the per-PID analysis in <code>LCD_extractCohort_callPerPID</code> , otherwise, no cutoffs are used.
<code>in_cohort_LCDlist</code>	Optional, if not provided, the cohort-wide exposures are recalculated by calling <code>LCD_complex_cutoff</code>
<code>in_perPID_LCDlist</code>	Optional, if not provided, the per sample exposures are recalculated by calling <code>LCD_complex_cutoff_perPID</code>
<code>addSigs_cohort_cutoff</code>	Numeric value for a cutoff: signatures which are detected in a fraction of the samples of the cohort greater than this cutoff are kept for the consensus set of signatures
<code>addSigs_perPID_cutoff</code>	Numeric value for a cutoff: signatures which are detected in one sample with exposure greater than this cutoff are kept for the consensus set of signatures
<code>addSigs_relAbs_cutoff</code>	Numeric value for a cutoff: signatures which are detected with at least this fraction of all variants cohort wide are kept for the consensus set of signatures
<code>keep.unassigned</code>	Boolean, if TRUE the exposures from the signatures which don't fulfill the criteria to be kept will be added and stored in the exposures as "unassigned", otherwise the exposures are rescaled.
<code>keep.all.cohort.sigs</code>	If TRUE (default), all signatures extracted cohort wide are kept, if FALSE, the function reevaluates whether the signatures extracted cohort wide still fulfill their criteria (i.e. exposures > cutoff) after perPID extraction.

Value

A list with entries:

- `exposures`: The exposures H , a numeric data frame with l rows and m columns, l being the number of signatures and m being the number of samples
- `norm_exposures`: The normalized exposures H , a numeric data frame with l rows and m columns, l being the number of signatures and m being the number of samples
- `signatures`: The reduced signatures that have exposures bigger than `in_cutoff`
- `choice`: Index vector of the reduced signatures in the input signatures
- `order`: Order vector of the signatures by exposure
- `residual_catalogue`: Numerical data frame (matrix) of the difference between fit (product of signatures and exposures) and input mutational catalogue
- `rss`: Residual sum of squares (i.e. sum of squares of the residual catalogue)
- `cosDist_fit_orig_per_matrix`: Cosine distance between the fit (product of signatures and exposures) and input mutational catalogue computed after putting the matrix into vector format (i.e. one scalar product for the whole matrix)
- `cosDist_fit_orig_per_col`: Cosine distance between the fit (product of signatures and exposures) and input mutational catalogue computed per column (i.e. per sample, i.e. as many scalar products as there are samples in the cohort)

- `sum_ind`: Decreasing order of mutational loads based on the input mutational catalogue
- `out_sig_ind_df`: Data frame of the type `signature_indices_df`, i.e. indicating name, function and meta-information of the signatures. Default is NULL, non-NULL only if `in_sig_ind_df` is non-NULL.
- `aggregate_exposures_list`: List of exposure data frames aggregated over different categories. Default is NULL, non-NULL only if `in_sig_ind_df` and `in_cat_list` are non-NULL and if the categories specified in `in_cat_list` are among the column names of `in_sig_ind_df`.

See Also

[LCD](#)

[aggregate_exposures_by_category](#)

[lsei](#)

Examples

```
NULL
```

LCD_SMC

CD stratification analysis

Description

CD stratification analysis

Usage

```
LCD_SMC(in_mutation_sub_catalogue_list, in_signatures_df, in_F_df = NULL)
```

Arguments

`in_mutation_sub_catalogue_list`

A list of s stratified mutational catalogues V_i (numeric data frames) with n rows and m columns each, n being the number of features and m being the number of samples. This list is naturally provided in [run_SMC](#).

`in_signatures_df`

A numeric data frame W with n rows and l columns, n being the number of features and l being the number of signatures

`in_F_df`

Default NULL

Value

Returns a list with all exposures and the stratified ones

logLikelihood	<i>Compute a loglikelihood ratio test</i>
---------------	---

Description

Compute a likelihood ratio test based on the loglikelihoods of the residuals of two different models of the same data.

Usage

```
logLikelihood(
  in_1,
  in_2,
  df_1 = NULL,
  df_2 = NULL,
  in_pdf = NULL,
  verbose = FALSE
)
```

Arguments

<code>in_1</code>	Residuals of model 1 of the input data.
<code>in_2</code>	Residuals of model 2 of the input data.
<code>df_1</code>	Degrees of freedom of the input model 1. If either <code>df_1</code> or <code>df_2</code> is NULL, the difference between the degrees of freedom of the two models is assumed to be 1.
<code>df_2</code>	Degrees of freedom of the input model 2. If either <code>df_1</code> or <code>df_2</code> is NULL, the difference between the degrees of freedom of the two models is assumed to be 1.
<code>in_pdf</code>	Probability distribution function, passed on to computeLogLik , if NULL a normal distribution is used.
<code>verbose</code>	Verbose if <code>in_verbose=1</code>

Value

A list with entries

- `statistic`: The test statistic
- `delta_df`: The difference in degrees of freedom between input model 1 and 2
- `p.value`: p value of the statistical test.

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
data(sigs)
data(cutoffs)
word_length <- 3
temp_list <- create_mutation_catalogue_from_df(
  lymphoma_test_df, this_seqnames.field = "CHROM",
```



```

    this_start.field = "POS",this_end.field = "POS",
    this_PID.field = "PID",this_subgroup.field = "SUBGROUP",
    this_refGenome = BSgenome.Hsapiens.UCSC.hg19,
    this_wordLength = word_length)
lymphoma_catalogue_df <- temp_list$matrix
lymphoma_PIDs <- colnames(lymphoma_catalogue_df)
current_sig_df <- AlexCosmicValid_sig_df
current_sigInd_df <- AlexCosmicValid_sigInd_df
current_cutoff_vector <- cutoffCosmicValid_rel_df[6, ]
iniLCDList <- LCD_complex_cutoff(
  in_mutation_catalogue_df = lymphoma_catalogue_df[, 1, drop = FALSE],
  in_signatures_df = current_sig_df,
  in_cutoff_vector = current_cutoff_vector,
  in_method = "relative", in_rescale = TRUE,
  in_sig_ind_df = current_sigInd_df)
current_sig_df <- AlexCosmicValid_sig_df[, -9]
current_sigInd_df <- AlexCosmicValid_sigInd_df[-9,]
current_cutoff_vector <- cutoffCosmicValid_rel_df[6, -9]
redLCDList <- LCD_complex_cutoff(
  in_mutation_catalogue_df = lymphoma_catalogue_df[, 1, drop = FALSE],
  in_signatures_df = current_sig_df,
  in_cutoff_vector = current_cutoff_vector,
  in_method = "relative", in_rescale = TRUE,
  in_sig_ind_df = current_sigInd_df)
logLikelihood(iniLCDList, redLCDList)

```

lymphomaNature2013_mutCat_df

Example mutational catalog for the SNV vignette

Description

lymphomaNature2013_mutCat_df: A data frame in the format of a SNV mutation catalog. The mutational catalog contains SNV variants from the lymphoma_Nature2013_raw_df data. Mutational catalog was created with create_mutation_catalogue_from_df function.

Usage

```
data(lymphomaNature2013_mutCat_df)
```

Value

A data frame in the layout of a SNV mutational catalog

References

ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/somatic_mutation_data/Lymphoma%20B-cell/Lymphoma%20B-cell_clean_somatic_mutations_for_signature_analysis.txt

Examples

```
data(LymphomaNature2013_mutCat_df)
head(LymphomaNature2013_mutCat_df)
dim(LymphomaNature2013_mutCat_df)
```

```
makeVRangesFromDataFrame
```

Construct a VRanges Object from a data frame

Description

In this package, big data frames are generated from cohort wide vcf-like files. This function constructs a VRanges object from such a data frame by using [makeGRangesFromDataFrame](#) from the package [GenomicRanges](#)

Usage

```
makeVRangesFromDataFrame(
  in_df,
  in_keep.extra.columns = TRUE,
  in_seqinfo = NULL,
  in_seqnames.field = "X.CHROM",
  in_start.field = "POS",
  in_end.field = "POS",
  in_PID.field = "PID",
  in_subgroup.field = "subgroup",
  in_strand.field = "strand",
  verbose_flag = 1
)
```

Arguments

<code>in_df</code>	A big dataframe constructed from a vcf-like file of a whole cohort. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or user defined columns. One of these can carry a PID (patient or sample identifier) and one can carry subgroup information.
<code>in_keep.extra.columns</code>	<code>in_seqinfo</code> Argument passed on to makeGRangesFromDataFrame
<code>in_seqinfo</code>	A seqInfo object, referring to the reference genome used. Argument passed on to makeGRangesFromDataFrame
<code>in_seqnames.field</code>	Indicates the name of the column in which the chromosome is encoded
<code>in_start.field</code>	Indicates the name of the column in which the start coordinate is encoded
<code>in_end.field</code>	Indicates the name of the column in which the end coordinate is encoded
<code>in_PID.field</code>	Indicates the name of the column in which the PID (patient or sample identifier) is encoded
<code>in_subgroup.field</code>	Indicates the name of the column in which the subgroup information is encoded

`in_strand.field` Indicates the name of the column in which the strandedness is encoded

`verbose_flag` Verbose if 1

Value

The constructed VRanges object

See Also

[makeGRangesFromDataFrame](#)

Examples

```
data(lymphoma_test)
temp_vr <- makeVRangesFromDataFrame(lymphoma_test_df,
                                   in_seqnames.field="CHROM",
                                   in_subgroup.field="SUBGROUP",
                                   verbose_flag=1)
```

make_catalogue_strata_df

Group strata from different stratification axes

Description

For a comparison of the strata from different orthogonal stratification axes, i.e. orthogonal SMCs, the strata have to be grouped and reformatted. This function does this task for the comparison by cosine similarity of mutational catalogues. Output of this function is the basis for applying [make_comparison_matrix](#). It is called by the wrapper function [run_comparison_catalogues](#).

Usage

```
make_catalogue_strata_df(
  in_stratification_lists_list,
  in_additional_stratum = NULL
)
```

Arguments

`in_stratification_lists_list`
List of lists with entries from different (orthogonal) stratification axes or SMCs

`in_additional_stratum`
Include an additionally supplied stratum in comparison in non-NULL.

Value

A list with entries `strata_df`, `number_of_SMCs`, `number_of_strata`.

- `strata_df`: Pasted numerical data frame of all strata (these are going to be compared e.g. by [make_comparison_matrix](#)).
- `number_of_SMCs`: Number of orthogonal stratifications in `in_stratification_lists_list` and additional ones.
- `number_of_strata`: Cumulative number of strata (sum over the numbers of strata of the different stratifications in `in_stratification_lists_list`) and additional ones.

See Also

[plot_strata](#)

[make_comparison_matrix](#)

[run_comparison_catalogues](#)

Examples

NULL

`make_comparison_matrix`

Compute a similarity matrix for different strata

Description

Compute and plot a similarity matrix for different strata from different stratification axes together. First, [compare_sets](#) is called on `in_strata_df` with itself, yielding a distance matrix (a numerical data frame) `dist_df` of the strata. The corresponding similarity matrix `1-dist_df` is then passed to [corrplot](#).

Usage

```
make_comparison_matrix(
  in_strata_df,
  output_path = NULL,
  in_nrect = 5,
  in_attribute = "",
  in_palette = NULL
)
```

Arguments

<code>in_strata_df</code>	Numerical data frame of all strata to be compared.
<code>output_path</code>	Path to directory where the results, especially the figure produced by corrplot is going to be stored.
<code>in_nrect</code>	Number of clusters in the clustering procedure provided by corrplot
<code>in_attribute</code>	Additional string for the file name where the figure produced by corrplot is going to be stored.
<code>in_palette</code>	Colour palette for the matrix

Value

The comparison matrix of cosine similarities.

See Also

[compare_SMCs](#)

Examples

```
data(sigs)
make_comparison_matrix(
  AlexCosmicValid_sig_df, in_nrect=9,
  in_palette=colorRampPalette(c("blue", "green", "red"))(n=100))
```

make_strata_df

Group strata from different stratification axes

Description

For a comparison of the strata from different orthogonal stratification axes, i.e. orthogonal SMCs, the strata have to be grouped and reformatted. This function does this task for the comparison by cosine similarity of signature exposures. Output of this function is the basis for applying [plot_strata](#) and [make_comparison_matrix](#). It is called by the wrapper functions [compare_SMCs](#), [run_plot_strata_general](#) or [run_comparison_general](#).

Usage

```
make_strata_df(
  in_stratification_lists_list,
  in_remove_signature_ind = NULL,
  in_additional_stratum = NULL
)
```

Arguments

`in_stratification_lists_list`

List of lists with entries from different (orthogonal) stratification axes or SMCs

`in_remove_signature_ind`

Omit one of the signatures in `in_signatures_ind_df` for the comparison if non-NULL. The parameter specifies the index of the signature to be removed.

`in_additional_stratum`

Include an additionally supplied stratum in comparison in non-NULL.

Value

A list with entries `strata_df`, `number_of_SMCs`, `number_of_strata`.

- `strata_df`: Pasted numerical data frame of all strata (these are going to be compared e.g. by [make_comparison_matrix](#)).
- `number_of_SMCs`: Number of orthogonal stratifications in `in_stratification_lists_list` and additional ones.

- `number_of_strata`: Cumulative number of strata (sum over the numbers of strata of the different stratifications in `in_stratification_lists_list`) and additional ones.

See Also

[plot_strata](#)
[make_comparison_matrix](#)
[compare_SMCs](#)
[run_plot_strata_general](#)
[run_comparison_general](#)

Examples

NULL

<code>make_subgroups_df</code>	<i>Make a custom data structure for subgroups</i>
--------------------------------	---

Description

Creates a data frame carrying the subgroup information and the order in which the PIDs have to be displayed. Calls [aggregate](#) on `in_vcf_like_df`.

Usage

```

make_subgroups_df(
  in_vcf_like_df,
  in_exposures_df = NULL,
  in_palette = NULL,
  in_subgroup.field = "SUBGROUP",
  in_PID.field = "PID",
  in_verbose = FALSE
)

```

Arguments

<code>in_vcf_like_df</code>	vcf-like data frame with point mutation calls
<code>in_exposures_df</code>	Data frame with the signature exposures
<code>in_palette</code>	Palette for colour attribution to the subgroups if non-NULL
<code>in_subgroup.field</code>	String indicating which column of <code>in_vcf_like_df</code> carries the subgroup information
<code>in_PID.field</code>	String indicating which column of <code>in_vcf_like_df</code> and of <code>in_exposures_df</code> carries the PID information
<code>in_verbose</code>	Whether verbose or not.

Value

subgroups_df: A data frame carrying the subgroup and rank information.

See Also

[aggregate](#)

Examples

```
data(lymphoma_test)
data(lymphoma_cohort_LCD_results)
choice_ind <- (names(lymphoma_Nature2013_COSMIC_cutoff_exposures_df)
              %in% unique(lymphoma_test_df$PID))
lymphoma_test_exposures_df <-
  lymphoma_Nature2013_COSMIC_cutoff_exposures_df[,choice_ind]
make_subgroups_df(lymphoma_test_df, lymphoma_test_exposures_df)
```

melt_exposures

Generically melts exposure data frames

Description

Melt an exposure data frame with signatures as ID variables.

Usage

```
melt_exposures(in_df)
```

Arguments

in_df Numeric data frame with exposures.

Value

A data frame with the molten exposures.

Examples

```
NULL
```

merge_exposures	<i>Merge exposure data frames</i>
-----------------	-----------------------------------

Description

Merges with the special feature of preserving the signatures and signature order.

Usage

```
merge_exposures(in_exposures_list, in_signatures_df)
```

Arguments

`in_exposures_list`

List of data frames (carrying information on exposures).

`in_signatures_df`

Data frame W in which the columns represent the signatures.

Value

A data frame with the merged exposures.

Examples

```
NULL
```

MutCat_indel_df	<i>Example mutational catalog for the Indel vignette</i>
-----------------	--

Description

MutCat_indel_df: A data frame in the format of a mutation catalog. The mutational catalog contains Indel variants from the GenomeOfNL_raw data. Variants were random sampled for 15 artificial patient for the purpose to have a Indel mutational catalog and have to show the functionality of the package. The results of the mutational catalog should not be interpreted for they biological relevance. Mutational catalog was created with `create_indel_mutation_catalogue_from_df` function.

Usage

```
data(GenomeOfNL_MutCat)
```

Value

A data fame in the layout of a Indel mutational catalog

References

Mutational catalog created form release version 5 of the Genome of NL https://www.nlgenome.nl/menu/main/app-go-nl/?page_id=9

Examples

```
data(GenomeOfN1_MutCat)
head(MutCat_indel_df)
dim(MutCat_indel_df)
```

```
normalizeMotifs_otherRownames
```

Normalize Somatic Motifs with different rownames

Description

This is a wrapper function to [normalizeMotifs](#). The rownames are first transformed to fit the convention of the [SomaticSignatures](#) package and then passed on to the above mentioned function.

Usage

```
normalizeMotifs_otherRownames(in_matrix, in_norms, adjust_counts = TRUE)
```

Arguments

`in_matrix`, `in_norms`

Arguments to [normalizeMotifs](#)

`adjust_counts` Whether to rescale the counts after adaption or not. Default is true.

Value

The matrix returned by [normalizeMotifs](#), but with rownames transformed back to the convention of the input

Examples

```
NULL
```

```
normalize_df_per_dim Useful functions on data frames
```

Description

`normalize_df_per_dim`: Normalization is carried out by dividing by `rowSums` or `colSums`; for rows with `rowSums=0` or columns with `colSums=0`, the normalization is left out.

`average_over_present`: If averaging over columns, zero rows (i.e. those with `rowSums=0`) are left out, if averaging over rows, zero columns (i.e. those with `colSums=0`) are left out.

`sd_over_present`: If computing the standard deviation over columns, zero rows (i.e. those with `rowSums=0`) are left out, if computing the standard deviation over rows, zero columns (i.e. those with `colSums=0`) are left out.

`stderrmean_over_present`: If computing the standard error of the mean over columns, zero rows (i.e. those with `rowSums=0`) are left out, if computing the standard error of the mean over rows, zero columns (i.e. those with `colSums=0`) are left out. Uses the function [stderrmean](#)

Usage

```
normalize_df_per_dim(in_df, in_dimension)

average_over_present(in_df, in_dimension)

sd_over_present(in_df, in_dimension)

stderrmean_over_present(in_df, in_dimension)
```

Arguments

```
in_df          Data frame to be normalized
in_dimension   Dimension along which the operation will be carried out
```

Value

The normalized numerical data frame (normalize_df_per_dim)
 A vector of the means (average_over_present)
 A vector of the standard deviations (sd_over_present)
 A vector of the standard errors of the mean (stderrmean_over_present)

See Also

[stderrmean](#)

Examples

```
test_df <- data.frame(matrix(c(1,2,3,0,5,2,3,4,0,6,0,0,0,0,4,5,6,0,7),
                             ncol=4))
## 1. Normalize over rows:
normalize_df_per_dim(test_df,1)
## 2. Normalize over columns:
normalize_df_per_dim(test_df,2)

test_df <- data.frame(matrix(c(1,2,3,0,5,2,3,4,0,6,0,0,0,0,4,5,6,0,7),
                             ncol=4))
## 1. Average over non-zero rows:
average_over_present(test_df,1)
## 2. Average over non-zero columns:
average_over_present(test_df,2)

test_df <- data.frame(matrix(c(1,2,3,0,5,2,3,4,0,6,0,0,0,0,4,5,6,0,7),
                             ncol=4))
## 1. Compute standard deviation over non-zero rows:
sd_over_present(test_df,1)
## 2. Compute standard deviation over non-zero columns:
sd_over_present(test_df,2)

test_df <- data.frame(matrix(c(1,2,3,0,5,2,3,4,0,6,0,0,0,0,4,5,6,0,7),
                             ncol=4))
## 1. Compute standard deviation over non-zero rows:
stderrmean_over_present(test_df,1)
## 2. Compute standard deviation over non-zero columns:
stderrmean_over_present(test_df,2)
```

plotExchangeSpectra *Plot the spectra of nucleotide exchanges*

Description

Plots the spectra of nucleotide exchanges in their triplet contexts. If several columns are present in the input data frame, the spectra are plotted for every column separately.

Usage

```
plotExchangeSpectra(  
  in_catalogue_df,  
  in_colour_vector = NULL,  
  in_show_triplets = FALSE,  
  in_show_axis_title = FALSE,  
  in_scales = "free_x",  
  in_refLine = NULL,  
  in_refAlpha = 0.5,  
  in_background = NULL  
)
```

Arguments

`in_catalogue_df` Numerical data frame encoding the exchange spectra to be displayed, either a mutational catalogue V or a signatures matrix W .

`in_colour_vector` Specifies the colours of the 6 nucleotide exchanges if non-null.

`in_show_triplets` Whether or not to show the triplets on the x-axis

`in_show_axis_title` Whether or not to show the name of the y-axis

`in_scales` Argument passed on to [facet_grid](#)

`in_refLine` If non-null, value on the y-axis at which a horizontal line is to be drawn

`in_refAlpha` Transparency of the horizontal line if it is to be drawn

`in_background` Option to provide a background theme, e.g. [theme_grey](#)

Value

The generated barplot - a ggplot2 plot

See Also

[geom_bar](#)
[facet_grid](#)

Examples

```
NULL
```

 plotExchangeSpectra_indel

Plot the spectra of nucleotide exchanges of INDELS

Description

Plots the spectra of nucleotides in their triplet contexts. If several columns are present in the input data frame, the spectra are plotted for every column separately. The function is only suitable for a INDEL spectra and for SNV representation the function [plotExchangeSpectra](#) should be used.

Usage

```
plotExchangeSpectra_indel(
  in_catalogue_df,
  in_colour_vector = NULL,
  in_show_indel = FALSE,
  in_show_axis_title = FALSE,
  in_scales = "free_x",
  in_refLine = NULL,
  in_refAlpha = 0.5,
  in_background = NULL
)
```

Arguments

<code>in_catalogue_df</code>	Numerical data frame encoding the exchange spectra to be displayed, either a mutational catalogue V or a signatures matrix W
<code>in_colour_vector</code>	Specifies the colours of the INDELS if non-null
<code>in_show_indel</code>	Whether or not to show the INDEL names on the x-axis
<code>in_show_axis_title</code>	Whether or not to show the name of the y-axis
<code>in_scales</code>	Argument passed on to facet_grid
<code>in_refLine</code>	If non-null, value on the y-axis at which a horizontal line is to be drawn
<code>in_refAlpha</code>	Transparency of the horizontal line if it is to be drawn
<code>in_background</code>	Option to provide a background theme, e.g. theme_grey

Value

The generated barplot - a ggplot2 plot

Examples

```
data(sigs_pcawg)
plotExchangeSpectra_indel(PCAWG_SP_ID_sigs_df[,c(6,8)])
```

plotExposuresConfidence

Plot exposures including confidence intervals

Description

Plot the exposures to extracted signatures including confidence intervals computed e.g. by [variateExp](#).

Usage

```
plotExposuresConfidence(in_complete_df, in_subgroups_df, in_sigInd_df)
```

Arguments

`in_complete_df` Melted numeric input data frame e.g. as computed by [variateExp](#)
`in_subgroups_df` Data frame containing meta information on subgroup attribution of the samples in the cohort of interest.
`in_sigInd_df` Data frame with meta information on the signatures used in the analysis.

Value

The function doesn't return any value but plots instead.

Examples

```
NULL
```

plotExposuresConfidence_indel

Plot exposures including confidence intervals for exposures of SNVs and INDELS

Description

Plot the exposures to extracted signatures including the confidence intervals computed e.g. by [variateExp](#)

Usage

```
plotExposuresConfidence_indel(in_complete_df, in_subgroups_df, in_sigInd_df)
```

Arguments

`in_complete_df` Melted numeric input data frame e.g. as computed by [variateExp](#)
`in_subgroups_df` Data frame containing meta information on subgroup attribution of the samples in the cohort of interest.
`in_sigInd_df` Data frame with meta information on the signatures used in the analysis.

Value

The function returns a gtable object which can be plotted with `plot` or `grid.draw`

Examples

```
NULL
```

plot_exposures	<i>Plot the exposures of a cohort</i>
----------------	---------------------------------------

Description

`plot_exposures`: The exposures H, determined by NMF or by LCD, are displayed as a stacked barplot by calling

- `geom_bar` and optionally
- `geom_text`.

The x-axis displays the PIDs (patient identifier or sample), the y-axis the counts attributed to the different signatures with their respective colours per PID. Is called by `plot_relative_exposures`.

`plot_relative_exposures`: Plot the relative or normalized exposures of a cohort. This function first normalizes its input and then sends the normalized data to `plot_exposures`.

Usage

```
plot_exposures(
  in_exposures_df,
  in_signatures_ind_df,
  in_subgroups_df = NULL,
  in_sum_ind = NULL,
  in_subgroups.field = "subgroup",
  in_title = "",
  in_labels = TRUE,
  in_show_subgroups = TRUE,
  legend_height = 10
)
```

```
plot_relative_exposures(
  in_exposures_df,
  in_signatures_ind_df,
  in_subgroups_df,
  in_sum_ind = NULL,
  in_subgroups.field = "subgroup",
  in_title = "",
  in_labels = TRUE,
  in_show_subgroups = TRUE
)
```

Arguments

<code>in_exposures_df</code>	Numerical data frame encoding the exposures H, i.e. which signature contributes how much to which PID (patient identifier or sample).
<code>in_signatures_ind_df</code>	A data frame containing meta information about the signatures
<code>in_subgroups_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup
<code>in_sum_ind</code>	Index vector influencing the order in which the PIDs are going to be displayed
<code>in_subgroups.field</code>	String indicating the column name in <code>in_subgroups_df</code> to take the subgroup information from.
<code>in_title</code>	Title for the plot to be created.
<code>in_labels</code>	Flag, if TRUE the PIDs are displayed on the x-axis
<code>in_show_subgroups</code>	Flag, if TRUE then PIDs are grouped by subgroups
<code>legend_height</code>	How many signatures should be displayed in one column together at most.

Value

The generated barplot - a ggplot2 plot

See Also

[LCD](#)

[geom_bar](#)

[geom_text](#)

Examples

```
data(lymphoma_cohort_LCD_results)
plot_exposures(lymphoma_Nature2013_COSMIC_cutoff_exposures_df,
               chosen_signatures_indices_df,
               COSMIC_subgroups_df)
```

```
data(lymphoma_cohort_LCD_results)
plot_relative_exposures(lymphoma_Nature2013_COSMIC_cutoff_exposures_df,
                       chosen_signatures_indices_df,
                       COSMIC_subgroups_df)
```

plot_SMC

Plot results of the Stratification of a Mutational Catalogue

Description

Plot a big composite figure with 3 columns: in the left column the per-PID absolute exposures will be shown, in the middle column the per_PID relative or normalized exposures will be shown, in the right column the cohort-wide exposures are shown (averaged over PIDs).

Usage

```
plot_SMC(
  number_of_strata,
  output_path,
  decomposition_method,
  number_of_sigs,
  name_list,
  exposures_strata_list,
  this_signatures_ind_df,
  this_subgroups_df,
  in_strata_order_ind,
  exposures_both_rel_df_list,
  cohort_method_flag,
  fig_width = 1200,
  fig_height = 900,
  fig_type = "png",
  in_label_orientation = "turn",
  this_sum_ind = NULL
)
```

Arguments

number_of_strata	Number of strata as deduced from <code>link{SMC}</code>
output_path	Path to file where the results are going to be stored. If NULL, the results will be plotted to the running environment.
decomposition_method	String for the filename of the generated barplot.
number_of_sigs	Number of signatures
name_list	Names of the constructed strata.
exposures_strata_list	The list of s strata specific exposures H_i , all are numerical data frames with l rows and m columns, l being the number of signatures and m being the number of samples
this_signatures_ind_df	A data frame containing meta information about the signatures
this_subgroups_df	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup

in_strata_order_ind	Index vector defining reordering of the strata
exposures_both_rel_df_list	A list of s strata specific cohortwide (i.e. averaged over cohort) normalized exposures
cohort_method_flag	Either or several of c("all_PIDs", "cohort", "norm_PIDs"), representing alternative ways to average over the cohort.
fig_width	Width of the figure to be plotted
fig_height	Height of the figure to be plotted
fig_type	png or pdf
in_label_orientation	Whether or not to turn the labels on the x-axis.
this_sum_ind	Optional set of indices for reordering the PIDs

Value

The function doesn't return any value.

Examples

NULL

plot_strata	<i>Plot all strata from different stratification axes together</i>
-------------	--

Description

Plot the cohort wide signature exposures of all strata from different stratification axes together. Naturally called by [compare_SMCs](#).

Usage

```
plot_strata(
  in_strata_list,
  in_signatures_ind_df,
  output_path = NULL,
  in_attribute = ""
)
```

Arguments

in_strata_list	Data structure created by <code>make_strata_df</code> or <code>make_catalogue_strata_df</code> in which the strata from different orthogonal stratification axes are reorganized in a consistent structure.
in_signatures_ind_df	A data frame containing meta information about the signatures
output_path	Path to directory where the results, especially the figure produced, are going to be stored.
in_attribute	Additional string for the file name where the figure output is going to be stored.

Value

The function doesn't return any value.

See Also

[compare_SMCs](#)

Examples

NULL

read_entry	<i>Read a single vcf-like file into a single data frame</i>
------------	---

Description

Note: this function uses [read.csv](#) to read vcf-like files into data frames for single samples. As it uses [read.csv](#), the default value for `comment.char` is "" and not "#" as it would have been for [read.table](#).

Usage

```
read_entry(
  current_ind,
  in_list,
  header = TRUE,
  in_header = NULL,
  variant_type = "SNV",
  delete.char = NULL,
  ...
)
```

```
read_list(in_list, in_parallel = FALSE, header = TRUE, in_header = NULL, ...)
```

Arguments

<code>current_ind</code>	Index of the file to read from the list provided below.
<code>in_list</code>	List of paths to vcf-like file to be read. The list may be named.
<code>header</code>	Boolean whether a header information should be read (as in read.table)
<code>in_header</code>	Vector of column names to be substituted if non-NULL.
<code>variant_type</code>	Default is "SNV" and provides additional plausibility and checks, omitted if other string
<code>delete.char</code>	Character to be deleted, e.g. in order to discriminate between comment lines and header lines, if non-NULL
<code>...</code>	Parameters passed on to read.table
<code>in_parallel</code>	If multicore functionality is provided on a compute cluster, this option may be set to TRUE in order to enhance speed.

Value

A vcf-like data frame

A list with entries:

- `vcf_like_df_list`: List of the read data frames
- `readVcf_time`: Object of class `proc_time`, which stores the time needed for reading in the data

Examples

```
NULL
```

```
NULL
```

`relateSigs`*Make unique assignments between sets of signatures*

Description

Make unique assignments between a set of given signatures and a set of new signatures.

Usage

```
relateSigs(querySigs, subjectSigs)
```

Arguments

`querySigs` The signatures to compare to (given signatures).

`subjectSigs` The signatures to be compared (new signatures).

Value

A list of comparison vectors

See Also

[compare_sets](#)

[disambiguateVector](#)

Examples

```
NULL
```

repeat_df	<i>Create a data frame with default values</i>
-----------	--

Description

Create a data frame with default values

Usage

```
repeat_df(in_value, in_rows, in_cols)
```

Arguments

in_value	Default entry to be repeated in the data frame
in_rows, in_cols	Dimensions of the data frame to be created

Value

The created data frame

Examples

```
## 1. Initialize with numeric value:
repeat_df(1,2,3)
## 2. Initialize with NA value:
repeat_df(NA,3,2)
## 3. Initialize with character:
repeat_df("a",4,3)
```

round_precision	<i>Round to a defined precision</i>
-----------------	-------------------------------------

Description

This function is an extension with regard to the function `round` from base R as it allows not only digits as precision, but can also round to a user-specified precision. The interval in which the rounding operation is to be carried out also can be specified by the user (default is the unit interval). Alternatively, breaks can be provided.

Usage

```
round_precision(x, breaks = NULL, in_precision = 0.05, in_interval = c(0, 1))
```

Arguments

x	Vector to be rounded
breaks	The breaks used for rounding. Default NULL
in_precision	Precision default 0.05
in_interval	Interval needs to be larger than the precision value

Value

A list with two entries:

- values: the rounded vector
- breaks: the breaks used for rounding

Examples

NULL

run_annotate_vcf_pl *Wrapper function to annotate addition information*

Description

Wrapper function to the perl script `annotate_vcf.pl` which annotates data of a track stored in `file_B` (may be different formats) to called variants stored in a vcf-like `file_A`.

Usage

```
run_annotate_vcf_pl(
  in_data_file,
  in_anno_track_file,
  in_new_column_name,
  out_file,
  in_data_file_type = "custom",
  in_anno_track_file_type = "bed",
  in_data_CHROM.field = "CHROM",
  in_data_POS.field = "POS",
  in_data_END.field = "POS"
)
```

Arguments

`in_data_file` Path to the input vcf-like file to be annotated

`in_anno_track_file`
 Path to the input file containing the annotation track

`in_new_column_name`
 String indicating the name of the column to be created for annotation.

`out_file` Path where the created files can be stored.

`in_data_file_type`
 custom for vcf-like

`in_anno_track_file_type`
 Type of the file `in_anno_track_file` containing the annotation track.

`in_data_CHROM.field`
 String indicating which column of `in_data_file` contains the chromosome information.

`in_data_POS.field`
String indicating which column of `in_data_file` contains the position information.

`in_data_END.field`
String indicating which column of `in_data_file` contains the end information if regions are considered.

Value

Return zero if no problems occur.

Examples

NULL

`run_comparison_catalogues`
Compare all strata from different stratifications

Description

Compare all strata from different orthogonal stratification axes, i.e. orthogonal SMCs by cosine similarity of mutational catalogues. Function similar to [run_comparison_general](#). First calls

- [make_catalogue_strata_df](#), then
- [make_comparison_matrix](#)

Usage

```
run_comparison_catalogues(
  in_stratification_lists_list,
  output_path = NULL,
  in_nrect = 5,
  in_attribute = ""
)
```

Arguments

`in_stratification_lists_list` List of lists with entries from different (orthogonal) stratification axes or SMCs

`output_path` Path to directory where the results, especially the figure produced by [corrplot](#) is going to be stored.

`in_nrect` Number of clusters in the clustering procedure provided by [corrplot](#)

`in_attribute` Additional string for the file name where the figure produced by

Value

The comparison matrix of cosine similarities.

See Also

[make_comparison_matrix](#)
[run_comparison_general](#)

Examples

NULL

run_comparison_general

Compare all strata from different stratifications

Description

Compare all strata from different orthogonal stratification axes, i.e. orthogonal SMCs by cosine similarity of signature exposures. Function similar to [compare_SMCs](#), but without calling [plot_strata](#). First calls

- [make_strata_df](#), then
- [make_comparison_matrix](#)

Usage

```
run_comparison_general(
  in_stratification_lists_list,
  output_path = NULL,
  in_nrect = 5,
  in_attribute = "",
  in_remove_signature_ind = NULL,
  in_additional_stratum = NULL
)
```

Arguments

<code>in_stratification_lists_list</code>	List of lists with entries from different (orthogonal) stratification axes or SMCs
<code>output_path</code>	Path to directory where the results, especially the figure produced by corrplot is going to be stored.
<code>in_nrect</code>	Number of clusters in the clustering procedure provided by corrplot
<code>in_attribute</code>	Additional string for the file name where the figure produced by corrplot is going to be stored.
<code>in_remove_signature_ind</code>	Omit one of the signatures in <code>in_signatures_ind_df</code> for the comparison if non-NULL. The parameter specifies the index of the signature to be removed.
<code>in_additional_stratum</code>	Include an additionally supplied stratum in comparison in non-NULL.

Value

The comparison matrix of cosine similarities.

See Also

[make_comparison_matrix](#)

[compare_SMCs](#)

[run_comparison_catalogues](#)

Examples

NULL

run_kmer_frequency_correction

Provide comprehensive correction factors for kmer content

Description

This function is analogous to [normalizeMotifs](#). If an analysis of mutational signatures is performed on e.g. Whole Exome Sequencing (WES) data, the signatures and exposures have to be adapted to the potentially different kmer (trinucleotide) content of the target capture. The present function takes as arguments paths to the used reference genome and target capture file. It extracts the sequence of the target capture by calling bedtools getfasta on the system command prompt. run_kmer_frequency_normalization then calls a custom made perl script kmer_frequencies.pl also included in this package to count the occurrences of the triplets in both the whole reference genome and the created target capture sequence. These counts are used for normalization as in [normalizeMotifs](#). Note that [kmerFrequency](#) provides a solution to approximate kmer frequencies by random sampling. As opposed to that approach, the function described here deterministically counts all occurrences of the kmers in the respective genome.

Usage

```
run_kmer_frequency_correction(  
  in_ref_genome_fasta,  
  in_target_capture_bed,  
  in_word_length,  
  project_folder,  
  target_capture_fasta = "targetCapture.fa",  
  in_verbose = 1  
)
```

Arguments

in_ref_genome_fasta

Path to the reference genome fasta file used.

in_target_capture_bed

Path to a bed file containing the information on the used target capture. May also be a compressed bed.

in_word_length	Integer number defining the length of the features or motifs, e.g. 3 for triplets or 5 for pentamers
project_folder	Path where the created files, especially the fasta file with the sequence of the target capture and the count matrices, can be stored.
target_capture_fasta	Name of the fasta file of the target capture to be created if not yet existent.
in_verbose	Verbose if in_verbose=1

Value

A list with 2 entries:

- rel_cor: The correction factors after normalization as in [run_kmer_frequency_normalization](#)
- abs_cor: The correction factors without normalization.

See Also

[normalizeMotifs](#)

Examples

NULL

run_kmer_frequency_normalization

Provide normalized correction factors for kmer content

Description

This function is analogous to [normalizeMotifs](#). If an analysis of mutational signatures is performed on e.g. Whole Exome Sequencing (WES) data, the signatures and exposures have to be adapted to the potentially different kmer (trinucleotide) content of the target capture. The present function takes as arguments paths to the used reference genome and target capture file. It extracts the sequence of the target capture by calling bedtools getfasta on the system command prompt. run_kmer_frequency_normalization then calls a custom made perl script kmer_frequencies.pl also included in this package to count the occurrences of the triplets in both the whole reference genome and the created target capture sequence. These counts are used for normalization as in [normalizeMotifs](#). Note that [kmerFrequency](#) provides a solution to approximate kmer frequencies by random sampling. As opposed to that approach, the function described here deterministically counts all occurrences of the kmers in the respective genome.

Usage

```
run_kmer_frequency_normalization(  
  in_ref_genome_fasta,  
  in_target_capture_bed,  
  in_word_length,  
  project_folder,  
  in_verbose = 1  
)
```

Arguments

in_ref_genome_fasta	Path to the reference genome fasta file used.
in_target_capture_bed	Path to a bed file containing the information on the used target capture. May also be a compressed bed.
in_word_length	Integer number defining the length of the features or motifs, e.g. 3 for triplets or 5 for pentamers
project_folder	Path where the created files, especially the fasta file with the sequence of the target capture and the count matrices, can be stored.
in_verbose	Verbose if in_verbose=1

Value

A numeric vector with correction factors

See Also

[normalizeMotifs](#)

Examples

NULL

run_plot_strata_general

Wrapper function for plot_strata

Description

First calls

- `make_strata_df`, then
- `plot_strata`

Usage

```
run_plot_strata_general(  
  in_stratification_lists_list,  
  in_signatures_ind_df,  
  output_path = NULL,  
  in_attribute = "",  
  in_remove_signature_ind = NULL,  
  in_additional_stratum = NULL  
)
```

Arguments

in_stratification_lists_list	List of lists with entries from different (orthogonal) stratification axes or SMCs
in_signatures_ind_df	A data frame containing meta information about the signatures
output_path	Path to directory where the results, especially the figure produced by plot_strata is going to be stored.
in_attribute	Additional string for the file name where the figure produced by plot_strata is going to be stored.
in_remove_signature_ind	Omit one of the signatures in in_signatures_ind_df for the comparison if non-NULL. The parameter specifies the index of the signature to be removed.
in_additional_stratum	Include an additionally supplied stratum in comparison in non-NULL.

Value

The function doesn't return any value.

See Also

[plot_strata](#)

Examples

NULL

run_SMC

Wrapper function for the Stratification of a Mutational Catalogue

Description

[run_SMC](#) takes as input a big dataframe constructed from a vcf-like file of a whole cohort. This wrapper function calls custom functions to construct a mutational catalogue and stratify it according to categories indicated by a special column in the input dataframe:

- [create_mutation_catalogue_from_df](#)
- [adjust_number_of_columns_in_list_of_catalogues](#)

This stratification yields a collection of stratified mutational catalogues, these are reformatted and sent to the custom function [SMC](#) and thus indirectly to [LCD_SMC](#) to perform a signature analysis of the stratified mutational catalogues. The result is then handed over to [plot_SMC](#) for visualization.

Usage

```
run_SMC(
  my_table,
  this_signatures_df,
  this_signatures_ind_df,
  this_subgroups_df,
  column_name,
  refGenome,
  cohort_method_flag = "all_PIDs",
  in_strata_order_ind = seq_len(length(unique(my_table[, column_name]))),
  wordLength = 3,
  verbose_flag = 1,
  target_dir = NULL,
  strata_dir = NULL,
  output_path = NULL,
  in_all_exposures_df = NULL,
  in_rownames = c(),
  in_norms = NULL,
  in_label_orientation = "turn",
  this_sum_ind = NULL
)
```

Arguments

<code>my_table</code>	A big dataframe constructed from a vcf-like file of a whole cohort. The first columns are those of a standard vcf file, followed by an arbitrary number of custom or user defined columns. One of these must carry a PID (patient or sample identifier) and one must be the category used for stratification.
<code>this_signatures_df</code>	A numeric data frame W in with n rows and l columns, n being the number of features and l being the number of signatures
<code>this_signatures_ind_df</code>	A data frame containing meta information about the signatures
<code>this_subgroups_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup
<code>column_name</code>	Name of the column in <code>my_table</code> which is going to be used for stratification
<code>refGenome</code>	FaFile of the reference genome to extract the motif context of the variants in <code>my_table</code>
<code>cohort_method_flag</code>	Either or several of <code>c("all_PIDs", "cohort", "norm_PIDs")</code> , representing alternative ways to average over the cohort.
<code>in_strata_order_ind</code>	Index vector defining reordering of the strata
<code>wordLength</code>	Integer number defining the length of the features or motifs, e.g. 3 for triplets or 5 for pentamers
<code>verbose_flag</code>	Verbose if <code>verbose_flag=1</code>
<code>target_dir</code>	Path to directory where the results of the stratification procedure are going to be stored if non-NULL.

strata_dir	Path to directory where the mutational catalogues of the different strata are going to be stored if non-NULL
output_path	Path to directory where the results, especially the figures produced by plot_SMC are going to be stored.
in_all_exposures_df	Optional argument, if specified, H, i.e. the overall exposures without stratification, is set to equal in_all_exposures_df. This is equivalent to forcing the LCD_SMC procedure to use e.g. the exposures of a previously performed NMF decomposition.
in_rownames	Optional parameter to specify rownames of the mutational catalogue V i.e. the names of the features.
in_norms	If specified, vector of the correction factors for every motif due to differing trinucleotide content. If null, no correction is applied.
in_label_orientation	Whether or not to turn the labels on the x-axis.
this_sum_ind	Optional set of indices for reordering the PIDs

Value

A list with entries `exposures_list`, `catalogues_list`, `cohort` and `name_list`.

- `exposures_list`: The list of `s` strata specific exposures H_i , all are numerical data frames with `l` rows and `m` columns, `l` being the number of signatures and `m` being the number of samples
- `catalogues_list`: A list of `s` strata specific cohortwide (i.e. averaged over cohort) normalized exposures
- `cohort`: `subgroups_df` adjusted for plotting
- `name_list`: Names of the constructed strata.

See Also

[create_mutation_catalogue_from_df](#)
[normalizeMotifs_otherRownames](#)
[plot_SMC](#)

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(sigs)
data(lymphoma_test)
data(lymphoma_cohort_LCD_results)
strata_list <-
  cut_breaks_as_intervals(lymphoma_test_df$random_norm,
                          in_outlier_cutoffs=c(-4,4),
                          in_cutoff_ranges_list=list(c(-2.5,-1.5),
                                                      c(0.5,1.5)),
                          in_labels=c("small", "intermediate", "big"))
lymphoma_test_df$random_cat <- strata_list$category_vector
choice_ind <- (names(lymphoma_Nature2013_COSMIC_cutoff_exposures_df)
              %in% unique(lymphoma_test_df$PID))
lymphoma_test_exposures_df <-
  lymphoma_Nature2013_COSMIC_cutoff_exposures_df[, choice_ind]
temp_subgroups_df <- make_subgroups_df(lymphoma_test_df,
```

```
                                lymphoma_test_exposures_df)
mut_density_list <- run_SMC(lymphoma_test_df,
                            AlexCosmicValid_sig_df,
                            AlexCosmicValid_sigInd_df,
                            temp_subgroups_df,
                            column_name="random_cat",
                            refGenome=BSgenome.Hsapiens.UCSC.hg19,
                            cohort_method_flag="norm_PIDs",
                            in_rownames = rownames(AlexCosmicValid_sig_df))
```

shapiro_if_possible *Wrapper for Shapiro test but allow for all identical values*

Description

Wrapper for Shapiro test but allow for all identical values

Usage

```
shapiro_if_possible(in_vector)
```

Arguments

`in_vector` Numerical vector the Shapiro-Wilk test is computed on

Value

p-value of the Shapiro-Wilk test, zero if all entries in the input vector `in_vector` are identical.

See Also

[shapiro.test](#)

Examples

```
shapiro_if_possible(runif(100,min=2,max=4))
shapiro_if_possible(rnorm(100,mean=5,sd=3))
shapiro_if_possible(rep(4.3,100))
shapiro_if_possible(c("Hello", "World"))
```

Description

The numerical data of the mutational signatures published initially by Alexandrov et al. (Nature 2013) and Alexandrov et al., (BioRxiv 2018) is stored in data frames with endings `_sig_df`, the associated meta-information is stored in data frames with endings `_sigInd_df`. There are several instances of `_sig_df` and `_sigInd_df`, corresponding to results and data obtained at different times and with different raw data. There always is a one-to-one correspondence between a `_sig_df` and a `_sigInd_df`. The data frames of type `_sig_df` have as many rows as there are features, i.e. 96 if analyzing mutational signatures of SNVs in a triplet context, and as many columns as there are signatures. Data frames of type `_sigInd_df` have as many rows as there are signatures in the corresponding `_sig_df` and several columns:

- `sig`: signature name
- `index`: corresponding to the row index of the signature
- `colour`: colour for visualization in stacked barplots
- `process`: asserted biological process
- `cat.coarse`: categorization of the signatures according to the asserted biological processes at low level of detail
- `cat.medium`: categorization of the signatures according to the asserted biological processes at intermediate level of detail
- `cat.high`: categorization of the signatures according to the asserted biological processes at high level of detail
- `cat.putative`: categorization of the signatures according to the asserted biological processes based on clustering and inference

Please note, that categorization columns are only present for the data frames corresponding to the data from Alexandrov et al. (Nature 2013).

`AlexInitialArtif_sig_df`: Data frame of the signatures published initially by Alexandrov et al. (Nature 2013). There are 27 signatures which constitute the columns, 22 of which were validated by an orthogonal sequencing technology. These 22 are in the first 22 columns of the data frame. The column names are *A* pasted to the number of the signature, e.g. *A5*. The nonvalidated signatures have an additional letter in their naming convention: either *AR1 - AR3* or *AU1 - AU2*. The rownames are the features, i.e. an encoding of the nucleotide exchanges in their trinucleotide context, e.g. *C>A ACA*. In total there are 96 different features and therefore 96 rows when dealing with a trinucleotide context.

`AlexInitialArtif_sigInd_df`: Meta-information for `AlexInitialArtif_sig_df`

`AlexInitialValid_sig_df`: Data frame of only the validated signatures published initially by Alexandrov et al. (Nature 2013), corresponding to the first 22 columns of `AlexInitialArtif_sig_df`

`AlexInitialValid_sigInd_df`: Meta-information for `AlexInitialValid_sig_df`

`AlexCosmicValid_sig_df`: Data frame of the updated signatures list maintained by Ludmil Alexandrov at <https://cancer.sanger.ac.uk/cosmic/signatures>. The column names are *AC* pasted to the number of the signature, e.g. *AC5*. The naming convention for the rows is as described for `AlexInitialArtif_sig_df`.

`AlexCosmicValid_sigInd_df`: Meta-information for `AlexCosmicValid_sig_df`

AlexCosmicArtif_sig_df: Data frame of the updated signatures list maintained by Ludmil Alexandrov at <https://cancer.sanger.ac.uk/cosmic/signatures> and complemented by the artifact signatures from the initial publication, i.e. the last 5 columns of AlexInitialArtif_sig_df. The column names are AC pasted to the number of the signature, e.g. AC5. The naming convention for the rows is as described for AlexInitialArtif_sig_df.

AlexCosmicArtif_sigInd_df: Meta-information for AlexCosmicArtif_sig_df

Usage

```
data(sigs)
```

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

Source

AlexInitial: <ftp://ftp.sanger.ac.uk/pub/cancer/AlexandrovEtAl/signatures.txt>

AlexCosmic: https://cancer.sanger.ac.uk/cancergenome/assets/signatures_probabilities.txt

References

Alexandrov et al. (Nature 2013)

sigs_pcawg

Data for PCAWG SNV signatures (COSMIC v3), including artifacts PCAWG_SP_SBS_sigs_Artif_df: Data frame of the signatures published by Alexandrov et al. (Biorxiv 2013) which were decomposed with the method SigProfiler. SNV signatures are labeled with SBS, single base signature. There are 67 signatures which constitute the columns, 47 of which were validated by a bayesian NFM method, SignatureAnalyzer. Validated signatures are SBS1-SBS26,SBS28-SBS42 and SBS44. SBS7 is split up into 7 a/b/c and d. SBS10 and SBS17 are both split up into a and b. Resulting in a 47 validated signatures. Please note, unlike the paper by Alexandrov et al. (Biorxiv 2018) the data sets do not contain a SBS84 and SBS85 as not all were available to perform supervised signature analysis. In total there are 96 different features and therefore 96 rows when dealing with a trinucleotide context.

Description

PCAWG_SP_SBS_sigInd_Artif_df: Meta-information for PCAWG_SP_SBS_sigs_Artif_df

PCAWG_SP_SBS_sigs_Real_df: Data frame of only the validated signatures published by Alexandrov et al. (Biorxiv 2018), corresponding to the column 1-26, 28-42 and 44 of the PCAWG_SP_SBS_sigs_Artif_df data frame

PCAWG_SP_SBS_sigInd_Real_df: Meta-information for PCAWG_SP_SBS_sigs_Real_df

PCAWG_SP_ID_sigs_df: Data frame with Indel signatures published by Alexandrov et al. (Biorxiv 2018) which were decomposed with the method SigProfiler. There are 17 Signatures reported but as

supervised signatures are only valid for whole genome sequencing data analysis. In whole genome sequencing data the Indel signature ID15 was not described and thus is not part of this data set. In total 83 features are described. The categorization considers the size of the insertion and deletion, the motif, and the sequence context. Hereby the number of repetition or partial repetition of the motif is determined.

PCAWG_SP_ID_sigInd_df: Meta-information for PCAWG_SP_ID_sigs_df

Usage

```
data(sigs_pcastg)
```

Author(s)

Lea Jopp-Saile <huebschmann.daniel@gmail.com>

Source

PCAWG_SNV: <https://www.synapse.org/#!/Synapse:syn11738319>

PCAWG_INDEL: <https://cancer.sanger.ac.uk/cosmic/signatures/ID>

References

Alexandrov et al. (Biorxiv 2018)

SMC

Stratification of a Mutational Catalogue

Description

SMC takes a given collection of stratified mutational catalogues V_i , sends them to perform a mutational signatures decomposition by Linear Combination Decomposition (LCD) with the functions `LCD_SMC` with known signatures W . It subsequently performs some useful statistics and preparation for plotting with the function `plot_SMC`. SMC is naturally called by `run_SMC`.

Usage

```
SMC(  
  df_list,  
  this_signatures_df,  
  in_all_exposures_df,  
  number_of_strata,  
  number_of_sigs,  
  name_list,  
  this_subgroups_df,  
  mutation_catalogue_all_df,  
  cohort_method_flag,  
  in_verbose = 1  
)
```

Arguments

<code>df_list</code>	A list of s stratified mutational catalogues V_i (numeric data frames) with n rows and m columns each, n being the number of features and m being the number of samples. This list is naturally provided in run_SMC .
<code>this_signatures_df</code>	A numeric data frame W in with n rows and l columns, n being the number of features and l being the number of signatures
<code>in_all_exposures_df</code>	The overall exposures H without stratification, a numeric data frame with l rows and m columns, l being the number of signatures and m being the number of samples
<code>number_of_strata</code>	The length of the list <code>df_list</code>
<code>number_of_sigs</code>	The number of signatures used in the current decomposition.
<code>name_list</code>	A list of names of the different strata
<code>this_subgroups_df</code>	A data frame indicating which PID (patient or sample identifier) belongs to which subgroup
<code>mutation_catalogue_all_df</code>	The overall mutational catalogue V without stratification.
<code>cohort_method_flag</code>	Either or several of <code>c("all_PIDs", "cohort", "norm_PIDs")</code> , representing alternative ways to average over the cohort.
<code>in_verbose</code>	Verbose if <code>in_verbose=1</code>

Value

A list with entries `exposures_strata_list`, `exposures_both_rel_df_list`, `this_subgroups_df`, `subgroup_ind` and `decomposition_method`.

- `exposures_strata_list`: The list of s strata specific exposures H_i , all are numerical data frames with l rows and m columns, l being the number of signatures and m being the number of samples
- `exposures_both_rel_df_list`: A list of s strata specific cohortwide (i.e. averaged over cohort) normalized exposures
- `this_subgroups_df`: `subgroups_df` adjusted for plotting
- `subgroup_ind`: Index of the subgroups chosen and relevant for plotting.
- `decomposition_method`: String telling whether LCD or NMF was used, relevant only for handing over to [plot_SMC](#).

See Also

[run_SMC](#)
[plot_SMC](#)
[LCD_SMC](#)

Examples

NULL

SMC_perPID

*Run SMC at a per sample level***Description**

Run an SMC analysis (stratification of the mutational catalogue) at per sample / per-PID level, corresponding to a divide and conquer strategy. For every single PID, only those signatures actually present in this PID will be provided for the SMC analysis.

Usage

```
SMC_perPID(
  in_dfList,
  in_LCDlist,
  in_subgroups_df,
  in_save_plot = TRUE,
  in_save_dir = NULL,
  in_save_name = "KataegisSMCs.pdf",
  in_verbose_flag = 0,
  ...
)
```

Arguments

<code>in_dfList</code>	Named list of vcf-like data frames, one entry per sample/PID of a cohort.
<code>in_LCDlist</code>	Output of an LCD list performed on the above cohort, carrying notably information on the exposures (<code>in_LCDlist\$exposures</code>), the present signatures (<code>in_LCDlist\$signatures</code>) and meta information about the signatures (<code>in_LCDlist\$out_sig_ind_df</code>).
<code>in_subgroups_df</code>	Data frame with subgroup information about the PIDs in the above mentioned cohort.
<code>in_save_plot</code>	Boolean flag to indicate whether per-PID plots should be saved.
<code>in_save_dir</code>	If per-PID plots are to be saved, this is the path where to save them.
<code>in_save_name</code>	Suffix to be appended to the sample name to generate the name of the saved per-PID plots.
<code>in_verbose_flag</code>	Whether to run verbose (1) or not (0).
<code>...</code>	Data passed on to run_SMC .

Value

A list of lists. The top level is a named per-PID list, each entry is of type `SMClist` (cf. [run_SMC](#)).

Examples

```
NULL
```

`split_exposures_by_subgroups`*Split an exposures data frame by subgroups*

Description

If a cohort consists of different subgroups, this function enables to split the data frame storing the signature exposures into a list of data frames with signature exposures, one per subgroup. This functionality is needed for [stat_test_subgroups](#) and [stat_plot_subgroups](#)

Usage

```
split_exposures_by_subgroups(  
  in_exposures_df,  
  in_subgroups_df,  
  in_subgroups.field = "subgroup",  
  in_PID.field = "PID"  
)
```

Arguments

<code>in_exposures_df</code>	Numerical data frame of the exposures (i.e. contributions of the different signatures to the number of point mutations per PID)
<code>in_subgroups_df</code>	Data frame indicating which PID belongs to which subgroup
<code>in_subgroups.field</code>	Name indicating which column in <code>in_subgroups_df</code> contains the subgroup information
<code>in_PID.field</code>	Name indicating which column in <code>in_subgroups_df</code> contains the PID information

Value

List of data frames with the subgroup specific signature exposures.

See Also

[stat_test_subgroups](#)

[stat_plot_subgroups](#)

Examples

```
NULL
```

stat_plot_subgroups *Plot averaged signature exposures per subgroup*

Description

Plot one averaged signature exposure pattern per subgroup. Uses [split_exposures_by_subgroups](#).

Usage

```
stat_plot_subgroups(  
  in_exposures_df,  
  in_subgroups_df,  
  in_signatures_ind_df,  
  in_subgroups.field = "subgroup",  
  in_PID.field = "PID",  
  in_colour_vector = NULL  
)
```

Arguments

`in_exposures_df` Numerical data frame of the exposures (i.e. contributions of the different signatures to the number of point mutations per PID)

`in_subgroups_df` Data frame indicating which PID belongs to which subgroup

`in_signatures_ind_df` Data frame carrying additional information on the signatures

`in_subgroups.field` Name indicating which column in `in_subgroups_df` contains the subgroup information

`in_PID.field` Name indicating which column in `in_subgroups_df` contains the PID information

`in_colour_vector` If non-null, specifies the colours attributed to the subgroups

Value

The function doesn't return any value, it plots instead.

See Also

[split_exposures_by_subgroups](#)

Examples

```
NULL
```

stat_test_SMC

Apply statistical tests to a stratification (SMC)

Description

stat_test_SMC tests for enrichment or depletion in the different strata of a stratification of the mutational catalogue for every signature independently by applying Kruskal Wallis tests. For those signatures where the Kruskal Wallis test gives a significant p-value, pairwise posthoc tests are carried out by calling `kwAllPairsNemenyiTest`. Additionally all data is tested for normality by Shapiro Wilk tests, so that the user may apply ANOVA and pairwise posthoc t-test where allowed.

Usage

```
stat_test_SMC(in_strat_list, in_flag = "norm")
```

Arguments

`in_strat_list` A list with entries `exposures_list`, `catalogues_list`, `cohort` and `name_list` as in the output of `run_SMC`.

- `exposures_list`: The list of `s` strata specific exposures H_i , all are numerical data frames with `l` rows and `m` columns, `l` being the number of signatures and `m` being the number of samples
- `catalogues_list`: A list of `s` strata specific cohortwide (i.e. averaged over cohort) normalized exposures
- `cohort`: `subgroups_df` adjusted for plotting
- `name_list`: Names of the constructed strata.

`in_flag` If "norm", all tests are performed on normalized exposures, otherwise the absolute exposures are taken.

Value

A list with entries `kruskal_df`, `shapiro_df`, `kruskal_posthoc_list`,

- `kruskal_df`: A data frame containing results (statistic and p values) of the Kruskal Wallis tests (tests for enrichment or depletion in the different strata for every signature independently).
- `shapiro_df`: A data frame containing results (p values) of the Shapiro Wilk tests (tests for normal distribution in the different strata for every signature independently).
- `kruskal_posthoc_list`: A list of results of pairwise posthoc tests carried out for those signatures where the Kruskal Wallis test yielded a significant p-value (carried out by `kwAllPairsNemenyiTest`).

See Also

[run_SMC](#)
[kwAllPairsNemenyiTest](#)
[kruskal.test](#)
[shapiro_if_possible](#)
[shapiro.test](#)

Examples

NULL

stat_test_subgroups *Test for differences in average signature exposures between subgroups*

Description

Apply Kruskal-Wallis tests to detect differences in the signature exposures between different subgroups. Uses [split_exposures_by_subgroups](#). Algorithm analogous to [stat_test_SMC](#).

Usage

```
stat_test_subgroups(  
  in_exposures_df,  
  in_subgroups_df,  
  in_subgroups.field = "subgroup",  
  in_PID.field = "PID"  
)
```

Arguments

`in_exposures_df` Numerical data frame of the exposures (i.e. contributions of the different signatures to the number of point mutations per PID)

`in_subgroups_df` Data frame indicating which PID belongs to which subgroup

`in_subgroups.field` Name indicating which column in `in_subgroups_df` contains the subgroup information

`in_PID.field` Name indicating which column in `in_subgroups_df` contains the PID information

Value

A list with entries `kruskal_df`, `kruskal_posthoc_list`,

- `kruskal_df`: A data frame containing results (statistic and p values) of the Kruskal Wallis tests (tests for enrichment or depletion in the different strata for every signature independently).
- `kruskal_posthoc_list`: A list of results of pairwise posthoc tests carried out for those signatures where the Kruskal Wallis test yielded a significant p-value (carried out by [kwAllPairsNemenyiTest](#)).

See Also

[split_exposures_by_subgroups](#)

[stat_test_SMC](#)

[kwAllPairsNemenyiTest](#)

[kruskal.test](#)

Examples

NULL

stderrmean

Compute the standard error of the mean

Description

This function returns the standard deviation of an input numerical vector divided by the square root of the length of the input vector

Usage

stderrmean(x)

Arguments

x A numerical vector

Value

Standard deviation of an input numerical vector divided by the square root of the length of the input vector

Examples

```
A <- c(1,2,3)
sd(A)
stderrmean(A)
```

sum_over_list_of_df

Elementwise sum over a list of (numerical) data frames

Description

Elementwise sum over a list of (numerical) data frames

Usage

sum_over_list_of_df(in_df_list)

Arguments

in_df_list List of (numerical) data frames

Value

A numerical data frame with the same dimensions as the entries of in_df_list with elementwise sums

Examples

```
A <- data.frame(matrix(c(1,1,1,2,2,2),ncol=2))
B <- data.frame(matrix(c(3,3,3,4,4,4),ncol=2))
df_list <- list(A=A,B=B)
sum_over_list_of_df(df_list)
```

targetCapture_cor_factors

Correction factors for different target capture kits

Description

List of lists with correction factors for different target capture kits. The elements of the overall list are lists, every one carrying information for one target capture kit (and named after it). The elements of these sublists are 64 dimensional vectors with correction factors for all triplets. They were computed using counts of occurrence of the respective triplets in the target capture and in the reference genome and making ratios (either for the counts themselves as in `abs_cor` or for the relative occurrences in `rel_cor`). The information in this data structure may be used as input to [normalizeMotifs_otherRownames](#).

Usage

```
data(targetCapture_cor_factors)
```

Value

A list of lists of data frames

Author(s)

Daniel Huebschmann <huebschmann.daniel@gmail.com>

testSigs

Test for significance of alternative models cohort wide

Description

Wrapper function for [variateExpSingle](#) for application cohort wide.

Usage

```
testSigs(
  in_catalogue_df,
  in_sig_df,
  in_exposures_df,
  in_factor = 0,
  in_pdf = NULL
)
```

Arguments

in_catalogue_df	Input numerical data frame of the mutational catalog of the cohort to be analyzed
in_sig_df	Numerical data frame of the signatures used for analysis.
in_exposures_df	Input numerical data frame of the exposures computed for the cohort to be analyzed
in_factor	Deviation factor of the altered alternative model.
in_pdf	Probability distribution function, parameter passed on to <code>confIntExp</code> if NULL assumed to be normal distribution.

Value

Returns a data frame

Examples

NULL

test_exposureAffected *Test significance of association*

Description

Test significance of association between a vector of exposures and a selection of samples, e.g. those affected by mutations in a pathway as returned by `find_affected_PIDs`

Usage

```
test_exposureAffected(
  in_exposure_vector,
  in_affected_PIDs,
  in_mutation_label = NULL,
  in_exposure_label = NULL
)
```

Arguments

in_exposure_vector	Named vector of a phenotype (e.g. exposures to a specific signature)
in_affected_PIDs	Character vector of samples affected by some criterion, e.g. mutations in a pathway as returned by <code>find_affected_PIDs</code>
in_mutation_label	If non-NULL, prefix to the mutation status (x-axis label) in the produced boxplot
in_exposure_label	If non-NULL, prefix to the exposures (y-axis label) in the produced boxplot

Value

A list with entries:

- `current_kruskal`: Kruskal test object from testing phenotype against affection
- `current_boxplot`: Boxplot of phenotype against affection

Examples

NULL

`test_gene_list_in_exposures`

Test if mutated PIDs are enriched in signatures

Description

For all signatures found in a project, this function tests whether PIDs having mutations in a specified list of genes of interest have significantly higher exposures.

Usage

```
test_gene_list_in_exposures(
  in_gene_list,
  in_exposure_df,
  in_mut_table,
  in_gene.field = "GENE_short",
  in_p_cutoff = 0.05
)
```

Arguments

<code>in_gene_list</code>	List with genes of interest
<code>in_exposure_df</code>	Data frame with the signature exposures
<code>in_mut_table</code>	Data frame or table of mutations (derived from vcf-format)
<code>in_gene.field</code>	Name of the column in which the gene names are to be looked up
<code>in_p_cutoff</code>	Significance threshold

Value

A list with entries `pvals`, `exposure_df`, `number_of_mutated`,

- `pvals`: p-values of the t-tests performed on mutated vs. unmutated PIDs
- `exposure_df`: Transposed input exposures data frame with additional annotations for mutation status
- `number_of_mutated`: Number of PIDs carrying a mutation

Examples

NULL

`transform_rownames_R_to_MATLAB`*Change rownames from one naming convention to another*

Description

Rownames or names of the features used differ between the different contexts a signature analysis is carried out in. The function `transform_rownames_R_to_MATLAB` changes from the convention used in the YAPSA package to the one used by Alexandrov et al. in the MATLAB framework.

The function `transform_rownames_MATLAB_to_R` changes from the convention used in Alexandrov et al. in the MATLAB framework to the one used by the YAPSA package.

The function `transform_rownames_MATLAB_to_R` changes from the convention used in stored mutational catalogues by Alexandrov et al. to the one used by the YAPSA package.

The function `transform_rownames_YAPSA_to_deconstructSigs` changes from the convention used in the YAPSA package to the one used by the `deconstructSigs` package.

The function `transform_rownames_YAPSA_to_deconstructSigs` changes from the convention used in the `deconstructSigs` package to the one used by the YAPSA package.

Usage

```
transform_rownames_R_to_MATLAB(in_rownames, wordLength = 3)
```

```
transform_rownames_MATLAB_to_R(in_rownames, wordLength = 3)
```

```
transform_rownames_nature_to_R(in_rownames, wordLength = 3)
```

```
transform_rownames_YAPSA_to_deconstructSigs(in_rownames, wordLength = 3)
```

```
transform_rownames_deconstructSigs_to_YAPSA(in_rownames, wordLength = 3)
```

Arguments

`in_rownames` Character vector of input rownames

`wordLength` Size of the considered motif context

Value

A character vector of the translated rownames.

Examples

```
NULL
```

translate_to_hg19 *Translate chromosome names to the hg19 naming convention*

Description

translate_to_hg19: In hg19 naming convention, chromosome names start with the prefix *chr* and the gonosomes are called *X* and *Y*. If data analysis is performed e.g. with [BSgenome.Hsapiens.UCSC.hg19](#), this naming convention is needed. The inverse transform is done with [translate_to_1kG](#).

translate_to_1kG: In 1kG, i.e. 1000 genomes naming convention, chromosome names have no prefix *chr* and the gonosomes are called *23* for *X* and *24* for *Y*. If data analysis is performed e.g. with [hs37d5.fa](#), this naming convention is needed. The inverse transform is done with [translate_to_hg19](#).

Usage

```
translate_to_hg19(in_dat, in_CHROM.field = "CHROM", in_verbose = FALSE)
```

```
translate_to_1kG(in_dat, in_CHROM.field = "chr", in_verbose = FALSE)
```

Arguments

`in_dat` GRanges object, VRanges object or data frame which carries one column with chromosome information to be reformatted.

`in_CHROM.field` String indicating which column of `in_dat` carries the chromosome information

`in_verbose` Whether verbose or not.

Value

GRanges object, VRanges object or data frame identical to `in_dat`, but with the names in the chromosome column replaced (if dealing with data frames) or alternatively the seqlevels replaced (if dealing with GRanges or VRanges objects).

Examples

```
test_df <- data.frame(CHROM=c(1,2,23,24),POS=c(100,120000000,300000,25000),
                      dummy=c("a","b","c","d"))
hg19_df <- translate_to_hg19(test_df, in_CHROM.field = "CHROM")
hg19_df

test_df <- data.frame(CHROM=c(1,2,23,24),POS=c(100,120000000,300000,25000),
                      dummy=c("a","b","c","d"))
hg19_df <- translate_to_hg19(test_df, in_CHROM.field = "CHROM")
onekG_df <- translate_to_1kG(hg19_df, in_CHROM.field = "CHROM")
onekG_df
```

trellis_rainfall_plot *Create a rainfall plot in a trellis structure*

Description

A trellis is a plot structure which allows space optimized multi-panel multi track plots. This function uses the package **gtrellis** developed by Zuguang Gu, also available at <https://www.bioconductor.org/packages/release/bioc/html/gtrellis.html>. The graphics in the tracks within a gtrellis plot are mostly drawn with functions from the package **grid**. Note that for technical reasons, the column indicating the chromosome MUST have the name *chr* and be the first column in the data frame supplied to the gtrellis functions. Therefore reformatting is performed in this function before calling gtrellis functions.

Usage

```
trellis_rainfall_plot(
  in_rainfall_dat,
  in_point_size = unit(1, "mm"),
  in_rect_list = NULL,
  in_title = "",
  in_CHROM.field = "CHROM",
  in_POS.field = "POS",
  in_dist.field = "dist",
  in_col.field = "col"
)
```

Arguments

<code>in_rainfall_dat</code>	Data frame which has to contain at least columns for chromosome, position, intermutational distance and colour information
<code>in_point_size</code>	size of the points in the rainfall plot to be created has to be provided with appropriate units, e.g. <code>in_point_size=unit(0.5,"mm")</code>
<code>in_rect_list</code>	Optional argument, if present, will lead to highlighting of specified regions by coloured but transparent rectangles
<code>in_title</code>	Title in the figure to be created.
<code>in_CHROM.field</code>	String indicating which column of <code>in_rainfall_dat</code> carries the chromosome information
<code>in_POS.field</code>	String indicating which column of <code>in_rainfall_dat</code> carries the position information
<code>in_dist.field</code>	String indicating which column of <code>in_rainfall_dat</code> carries the intermutational distance information
<code>in_col.field</code>	String indicating which column of <code>in_rainfall_dat</code> carries the colour information encoding the nucleotide exchange

Value

The function doesn't return any value.

The function doesn't return any value.

See Also

[gtrellis_layout](#)
[add_track](#)
[grid.points](#)

Examples

```
data(lymphoma_test)
choice_PID <- "4121361"
PID_df <- subset(lymphoma_test_df, PID==choice_PID)
trellis_rainfall_plot(PID_df, in_point_size=unit(0.5, "mm"))
```

variateExp

Wrapper to compute confidence intervals for a cohort

Description

Wrapper function around [confIntExp](#), which is applied to every signature/sample pair in a cohort. The extracted upper and lower bounds of the confidence intervals are added to the input data which is reordered and melted in order to prepare for visualization with `ggplot2`.

Usage

```
variateExp(
  in_catalogue_df,
  in_sig_df,
  in_exposures_df,
  in_sigLevel = 0.05,
  in_delta = 0.4,
  in_pdf = NULL
)
```

Arguments

<code>in_catalogue_df</code>	Input numerical data frame of the mutational catalog of the cohort to be analyzed.
<code>in_sig_df</code>	Numerical data frame of the signatures used for analysis.
<code>in_exposures_df</code>	Input numerical data frame of the exposures computed for the cohort to be analyzed.
<code>in_sigLevel</code>	Significance level, parameter passed to confIntExp .
<code>in_delta</code>	Inflation parameter for the alternative model, parameter passed on to confIntExp
<code>in_pdf</code>	Probability distribution function, parameter passed on to confIntExp , if NULL assumed to be normal distribution.

Value

A melted data frame.

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
data(lymphoma_cohort_LCD_results)
data(sigs)
word_length <- 3
temp_list <- create_mutation_catalogue_from_df(
  lymphoma_test_df, this_seqnames.field = "CHROM",
  this_start.field = "POS", this_end.field = "POS",
  this_PID.field = "PID", this_subgroup.field = "SUBGROUP",
  this_refGenome = BSgenome.Hsapiens.UCSC.hg19,
  this_wordLength = word_length)
lymphoma_catalogue_df <- temp_list$matrix
lymphoma_PIDs <- colnames(lymphoma_catalogue_df)
data("lymphoma_cohort_LCD_results")
lymphoma_exposures_df <-
  lymphoma_Nature2013_COSMIC_cutoff_exposures_df[,lymphoma_PIDs]
lymphoma_sigs <- rownames(lymphoma_exposures_df)
lymphoma_sig_df <- AlexCosmicValid_sig_df[,lymphoma_sigs]
lymphoma_complete_df <- variateExp(in_catalogue_df = lymphoma_catalogue_df,
                                   in_sig_df = lymphoma_sig_df,
                                   in_exposures_df = lymphoma_exposures_df,
                                   in_sigLevel = 0.025, in_delta = 0.4)

head(lymphoma_complete_df)
lymphoma_complete_df$sample <-
  factor(lymphoma_complete_df$sample,
         levels = colnames(lymphoma_exposures_df)[
           order(colSums(lymphoma_exposures_df), decreasing = TRUE)])
sig_colour_vector <- c("black", AlexCosmicValid_sigInd_df$colour)
names(sig_colour_vector) <-
  c("total", as.character(AlexCosmicValid_sigInd_df$sig))
ggplot(data = lymphoma_complete_df,
        aes(x = sample, y = exposure, fill = sig)) +
  geom_bar(stat = "identity") +
  geom_errorbar(aes(ymin = lower, ymax = upper), width = 0.2) +
  facet_wrap(~sig, nrow = nrow(lymphoma_exposures_df) + 1) +
  theme_grey() +
  theme(panel.border = element_rect(fill = NA, colour = "black"),
        strip.background = element_rect(colour = "black"),
        legend.position = "none") +
  scale_fill_manual(values = sig_colour_vector)
```

variateExpSingle

Wrapper for the likelihood ratio test

Description

Application of the likelihood ratio test to mutational signatures, primarily for one single sample.

Usage

```
variateExpSingle(
  in_catalogue_vector,
  in_sig_df,
  in_exposure_vector,
  in_ind,
  in_factor = 1,
  in_pdf = NULL,
  verbose = FALSE
)
```

Arguments

<code>in_catalogue_vector</code>	Mutational catalog of the input sample.
<code>in_sig_df</code>	Data frame encoding the signatures used for the analysis.
<code>in_exposure_vector</code>	Exposure vector computed for the input sample.
<code>in_ind</code>	Index specifying which signature among <code>in_sig_df</code> is to be tested.
<code>in_factor</code>	Deviation factor of the altered alternative model.
<code>in_pdf</code>	Probability distribution function, parameter passed on to logLikelihood and later to computeLogLik
<code>verbose</code>	Verbose if <code>in_verbose=1</code>

Value

Returns a list

Examples

```
library(BSgenome.Hsapiens.UCSC.hg19)
data(lymphoma_test)
data(lymphoma_cohort_LCD_results)
data(sigs)
word_length <- 3
temp_list <- create_mutation_catalogue_from_df(
  lymphoma_test_df, this_seqnames.field = "CHROM",
  this_start.field = "POS", this_end.field = "POS",
  this_PID.field = "PID", this_subgroup.field = "SUBGROUP",
  this_refGenome = BSgenome.Hsapiens.UCSC.hg19,
  this_wordLength = word_length)
lymphoma_catalogue_df <- temp_list$matrix
lymphoma_PIDs <- colnames(lymphoma_catalogue_df)
data("lymphoma_cohort_LCD_results")
lymphoma_exposures_df <-
  lymphoma_Nature2013_COSMIC_cutoff_exposures_df[, lymphoma_PIDs]
lymphoma_sigs <- rownames(lymphoma_exposures_df)
lymphoma_sig_df <- AlexCosmicValid_sig_df[, lymphoma_sigs]
variateExpSingle(
  in_ind = 1,
  in_factor = 1.5,
  in_catalogue_vector = lymphoma_catalogue_df[, 1],
  in_sig_df = lymphoma_sig_df,
```

```
in_exposure_vector = lymphoma_exposures_df[, 1])
```

YAPSA

Generate R documentation from inline comments.

Description

Yet Another Package for mutational Signature analysis

Details

This package provides functions and routines useful in the analysis of mutational signatures (cf. L. Alexandrov et al., Nature 2013). In particular, functions to perform a signature analysis with known signatures (**LCD** = linear combination decomposition) and a signature analysis on stratified mutational catalogue (**run_SMC** = stratify mutational catalogue) are provided.

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