

Package ‘CIMICE’

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Type Package

Title CIMICE-R: (Markov) Chain Method to Infer Cancer Evolution

Version 1.14.0

Description CIMICE is a tool in the field of tumor phylogenetics and its goal is to build a Markov Chain (called Cancer Progression Markov Chain, CPMC) in order to model tumor subtypes evolution. The input of CIMICE is a Mutational Matrix, so a boolean matrix representing altered genes in a collection of samples. These samples are assumed to be obtained with single-cell DNA analysis techniques and the tool is specifically written to use the peculiarities of this data for the CMPC construction.

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Author Nicolò Rossi [aut, cre] (Lab. of Computational Biology and Bioinformatics, Department of Mathematics, Computer Science and Physics, University of Udine,
<https://orcid.org/0000-0002-6353-7396>)

Maintainer Nicolò Rossi <olocin.issor@gmail.com>

Contents

annotate_mutational_matrix	3
binary_radix_sort	4
build_subset_graph	4
build_topology_subset	5
chunk_reader	6
CIMICE	6
compact_dataset	7
computeDWNW	7
computeDWNW_aux	8
computeUPW	9
computeUPW_aux	10
compute_weights_default	10
corrplot_from_mutational_matrix	11
corrplot_genes	11
corrplot_samples	12
dataset_preprocessing	12
dataset_preprocessing_population	13
draw_ggraph	14
draw_networkD3	14
draw_visNetwork	15
example_dataset	15
example_dataset_withFreqs	16
finalize_generator	16
fix_clonal_genotype	17
format_labels	18
gene_mutations_hist	18
get_no_of_children	19
graph_non_transitive_subset_topology	20
make_dataset	20
make_generator_stub	21
make_labels	21
normalizeDWNW	22
normalizeUPW	23
perturb_dataset	23
plot_generator	24
prepare_generator_edge_set_command	25
prepare_labels	26
quick_run	27
read	27
read_CAPRI	28
read_CAPRIpop	28
read_CAPRIpop_string	29
read_CAPRI_string	29

<i>annotate_mutational_matrix</i>	3
read_MAF	30
read_matrix	31
remove_transitive_edges	31
sample_mutations_hist	32
select_genes_on_mutations	32
select_samples_on_mutations	33
set_generator_edges	34
simulate_generator	35
to_dot	36
update_df	36
Index	38

`annotate_mutational_matrix`
Add samples and genes names to a mutational matrix

Description

Given M mutational matrix, add samples as row names, and genes as column names. If there are repetitions in row names, these are solved by adding a sequential identifier to the names.

Usage

```
annotate_mutational_matrix(M, samples, genes)
```

Arguments

M	mutational matrix
samples	list of sample names
genes	list of gene names

Value

N with the set row and column names

Examples

```
require(Matrix)
genes <- c("A", "B", "C")
samples <- c("S1", "S2", "S2")
M <- Matrix(c(0,0,1,0,0,1,0,1,1), ncol=3, sparse=TRUE, byrow = TRUE)

annotate_mutational_matrix(M, samples, genes)
```

binary_radix_sort *Radix sort for a binary matrix*

Description

Sort the rows of a binary matrix in ascending order

Usage

```
binary_radix_sort(mat)
```

Arguments

mat a binary matrix (of 0 and 1)

Value

the sorted matrix

Examples

```
require(Matrix)
m <- Matrix(c(1,1,0,1,0,0,0,1,1), sparse = TRUE, ncol = 3)
binary_radix_sort(m)
```

build_subset_graph *Remove transitive edges and prepare graph*

Description

Create a graph from the "build_topology_subset" edge list, so that it respects the subset relation, omitting the transitive edges.

Usage

```
build_subset_graph(edges, labels)
```

Arguments

edges edge list, built from "build_topology_subset"
labels list of node labels, to be paired with the graph

Value

a graph with the subset topology, omitting transitive edges

Examples

```
require(dplyr)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
edges <- build_topology_subset(samples)
g <- build_subset_graph(edges, labels)
```

build_topology_subset *Compute subset relation as edge list*

Description

Create an edge list E representing the 'subset' relation for binary strings so that:

$$(A, B) \text{ in } E \iff \text{forall}(i) : A[i] - > B[i]$$

Usage

```
build_topology_subset(samples)
```

Arguments

samples input dataset (mutational matrix) as matrix

Value

the computed edge list

Examples

```
require(dplyr)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
build_topology_subset(samples)
```

chunk_reader	<i>Gradually read a file from disk</i>
--------------	--

Description

This function creates a reader to read a text file in batches (or chunks). It can be used for very large files that cannot fit in RAM.

Usage

```
chunk_reader(file_path)
```

Arguments

file_path path to large file

Value

a list-object containing the function ‘read’ to read lines from the given file, and ‘close’ to close the connection to the file stream.

Examples

```
# open connection to file
reader <- chunk_reader(
  system.file("extdata", "paac_jhu_2014_500.maf", package = "CIMICE", mustWork = TRUE)
)

while(TRUE){
  # read a chunk
  chunk <- reader$read(10)
  if(length(chunk) == 0){
    break
  }
  # --- process chunk ---
}
# close connection
reader$close()
```

CIMICE	<i>CIMICE Package</i>
--------	-----------------------

Description

R implementation of the CIMICE tool. CIMICE is a tool in the field of tumor phylogenetics and its goal is to build a Markov Chain (called Cancer Progression Markov Chain, CPMC) in order to model tumor subtypes evolution. The input of CIMICE is a Mutational Matrix, so a boolean matrix representing altered genes in a collection of samples. These samples are assumed to be obtained with single-cell DNA analysis techniques and the tool is specifically written to use the peculiarities of this data for the CMPC construction. See <https://github.com/redsnic/tumorEvolutionWithMarkovChains/tree/master/> for the original Java version of this tool.

Details

CIMICE-R: (Markov) Chain Method to Infer Cancer Evolution

Author(s)

Nicolò Rossi <olocin.issor@gmail.com>

compact_dataset	<i>Compact dataset rows</i>
-----------------	-----------------------------

Description

Count duplicate rows and compact the dataset (mutational). The column 'freq' will contain the counts for each row.

Usage

```
compact_dataset(mutmatrix)
```

Arguments

mutmatrix input dataset (mutational matrix)

Value

a list with matrix (the compacted dataset (mutational matrix)), counts (frequencies of genotypes) and row_names (comma separated string of sample IDs) fields

Examples

```
compact_dataset(example_dataset())
```

computeDWNW	<i>Down weights computation</i>
-------------	---------------------------------

Description

Computes the Down weights formula using a Dinamic Programming approach (starting call), see vignettes for further explanation.

Usage

```
computeDWNW(g, freqs, no.of.children, A, normUpWeights)
```

Arguments

g graph (a Directed Acyclic Graph)
freqs observed genotype frequencies
no.of.children number of children for each node
A adjacency matrix of G
normUpWeights normalized up weights as computed by normalizeUPW

Value

a vector containing the Up weights for each edge

Examples

```

require(dplyr)
require(igraph)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
# prepare adj matrix
A <- as.matrix(as_adj(g))
# pre-compute exiting edges from each node
no.of.children <- get_no_of_children(A,g)
upWeights <- computeUPW(g, freqs, no.of.children, A)
normUpWeights <- normalizeUPW(g, freqs, no.of.children, A, upWeights)
computeDWNW(g, freqs, no.of.children, A, normUpWeights)

```

computeDWNW_aux

Down weights computation (aux)

Description

Computes the Down weights formula using a Dinamic Programming approach (recursion), see vignettes for further explanation.

Usage

```
computeDWNW_aux(g, edge, freqs, no.of.children, A, normUpWeights)
```

Arguments

g graph (a Directed Acyclic Graph)
edge the currently considered edge
freqs observed genotype frequencies
no.of.children number of children for each node
A adjacency matrix of G
normUpWeights normalized up weights as computed by normalizeUPW

Value

a vector containing the Up weights for each edge

computeUPW	<i>Up weights computation</i>
------------	-------------------------------

Description

Computes the up weights formula using a Dinamic Programming approach (starting call), see vignettes for further explanation.

Usage

```
computeUPW(g, freqs, no.of.children, A)
```

Arguments

<code>g</code>	graph (a Directed Acyclic Graph)
<code>freqs</code>	observed genotype frequencies
<code>no.of.children</code>	number of children for each node
<code>A</code>	adjacency matrix of G

Value

a vector containing the Up weights for each edge

Examples

```
require(dplyr)
require(igraph)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
# prepare adj matrix
A <- as.matrix(as_adj(g))
# pre-compute exiting edges from each node
no.of.children <- get_no_of_children(A,g)
computeUPW(g, freqs, no.of.children, A)
```

computeUPW_aux	<i>Up weights computation (aux)</i>
----------------	-------------------------------------

Description

Computes the up weights formula using a Dinamic Programming approach (recursion), see vignettes for further explanation.

Usage

```
computeUPW_aux(g, edge, freqs, no.of.children, A)
```

Arguments

g	graph (a Directed Acyclic Graph)
edge	the currently considered edge
freqs	observed genotype frequencies
no.of.children	number of children for each node
A	adjacency matrix of G

Value

a vector containing the Up weights for each edge

compute_weights_default	<i>Compute default weights</i>
-------------------------	--------------------------------

Description

This procedure computes the weights for edges of a graph accordingly to CIMICE specification. (See vignettes for further explanations)

Usage

```
compute_weights_default(g, freqs)
```

Arguments

g	a graph (must be a DAG with no transitive edges)
freqs	observed frequencies of genotypes

Value

a graph with the computed weights

Examples

```
require(dplyr)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
compute_weights_default(g, freqs)
```

`corrplot_from_mutational_matrix`*Correlation plot from mutational matrix*

Description

Prepare correlation plot based on a mutational matrix

Usage

```
corrplot_from_mutational_matrix(mutmatrix)
```

Arguments

`mutmatrix` input dataset

Value

the computed correlation plot

Examples

```
corrplot_from_mutational_matrix(example_dataset())
```

`corrplot_genes`*Gene based correlation plot*

Description

Prepare a correlation plot computed from genes' perspective using a mutational matrix

Usage

```
corrplot_genes(mutmatrix)
```

Arguments

`mutmatrix` input dataset (mutational matrix)

Value

the computed correlation plot

Examples

```
corrplot_genes(example_dataset())
```

corrplot_samples *Sample based correlation plot*

Description

Prepare a correlation plot computed from samples' perspective using a mutational matrix

Usage

```
corrplot_samples(mutmatrix)
```

Arguments

mutmatrix input dataset (mutational matrix)

Value

the computed correlation plot

Examples

```
corrplot_samples(example_dataset())
```

dataset_preprocessing *Run CIMICE preprocessing*

Description

executes the preprocessing steps of CIMICE

Usage

```
dataset_preprocessing(dataset)
```

Arguments

dataset a mutational matrix as a (sparse) matrix

Details

Preprocessing steps:

- 1) dataset is compacted
- 2) genotype frequencies are computed
- 3) labels are prepared

Value

a list containing the mutational matrix ("samples"), the mutational frequencies of the genotypes ("freqs"), the node labels ("labels") and finally the gene names ("genes")

Examples

```
require(dplyr)
example_dataset() %>% dataset_preprocessing
```

dataset_preprocessing_population

Run CIMICE preprocessing for population format dataset

Description

executes the preprocessing steps of CIMICE

Usage

```
dataset_preprocessing_population(compactDataset)
```

Arguments

compactDataset

a list (matrix: a mutational matrix, counts: number of samples with given genotype). "counts" is normalized automatically.

Details

Preprocessing steps:

- 1) genotype frequencies are computed
- 2) labels are prepared

Value

a list containing the mutational matrix ("samples"), the mutational frequencies of the genotypes ("freqs"), the node labels ("labels") and finally the gene names ("genes")

Examples

```
require(dplyr)
example_dataset_withFreqs() %>% dataset_preprocessing_population
```

draw_ggraph *ggplot graph output*

Description

Draws the output graph using ggplot

Usage

```
draw_ggraph(out, digits = 4, ...)
```

Arguments

out the output object of CIMICE (es, from quick run)
digits precision for edges' weights
... other arguments for format_labels

Value

ggraph object representing g as described

Examples

```
draw_ggraph(quick_run(example_dataset()))
```

draw_networkD3 *NetworkD3 graph output*

Description

Draws the output graph using networkD3

Usage

```
draw_networkD3(out, ...)
```

Arguments

out the output object of CIMICE (es, from quick run)
... other arguments for format_labels

Value

networkD3 object representing g as described

Examples

```
draw_networkD3(quick_run(example_dataset()))
```

draw_visNetwork	<i>VisNetwork graph output (default)</i>
-----------------	--

Description

Draws the output graph using VisNetwork

Usage

```
draw_visNetwork(out, ...)
```

Arguments

out	the output object of CIMICE (es, from quick run)
...	other arguments for format_labels

Value

visNetwork object representing g as described

Examples

```
draw_visNetwork(quick_run(example_dataset()))
```

example_dataset	<i>Creates a simple example dataset</i>
-----------------	---

Description

Creates a simple example dataset

Usage

```
example_dataset()
```

Value

a simple mutational matrix

Examples

```
example_dataset()
```

```
example_dataset_withFreqs
```

Creates a simple example dataset with frequency column

Description

Creates a simple example dataset with frequency column

Usage

```
example_dataset_withFreqs()
```

Value

a simple mutational matrix

Examples

```
example_dataset_withFreqs()
```

```
finalize_generator
```

Finalize generator normalizing edge weights

Description

Checks if a generator can be normalized so that it actually is a Markov Chain

Usage

```
finalize_generator(generator)
```

Arguments

generator a generator

Value

A generator with edge weights that respect DTMC definition

Examples

```
require(dplyr)
```

```
example_dataset() %>%
  make_generator_stub() %>%
  set_generator_edges(
    list(
      "D", "A, D", 1 ,
      "A", "A, D", 1 ,
      "A, D", "A, C, D", 1 ,
```



```

    "A, D", "A, B, D", 1 ,
    "Clonal", "D", 1 ,
    "Clonal", "A", 1 ,
    "D", "D", 1 ,
    "A", "A", 1 ,
    "A, D", "A, D", 1 ,
    "A, C, D", "A, C, D", 1 ,
    "A, B, D", "A, B, D", 1 ,
    "Clonal", "Clonal", 1
  )) %>%
  finalize_generator

```

fix_clonal_genotype *Manage Clonal genotype in data*

Description

Fix the absence of the clonal genotype in the data (if needed)

Usage

```
fix_clonal_genotype(samples, freqs, labels, matching_samples)
```

Arguments

samples	input dataset (mutational matrix) as matrix
freqs	genotype frequencies (in the rows' order)
labels	list of gene names (in the columns' order)
matching_samples	list of sample names matching each genotype

Value

a named list containing the fixed "samples", "freqs" and "labels"

Examples

```

require(dplyr)

# compact
compactDataset <- compact_dataset(example_dataset())
samples <- compactDataset$matrix

# save genes' names
genes <- colnames(compactDataset$matrix)

# keep the information on frequencies for further analysis
freqs <- compactDataset$counts/sum(compactDataset$counts)

# prepare node labels listing the mutated genes for each node
labels <- prepare_labels(samples, genes)
if( is.null(compactDataset$row_names) ){

```

```

    compactedDataset$row_names <- rownames(compactedDataset$matrix)
  }
  matching_samples <- compactedDataset$row_names
  # matching_samples
  matching_samples

  # fix Clonal genotype absence, if needed
  fix <- fix_clonal_genotype(samples, freqs, labels, matching_samples)

```

format_labels	<i>Format labels for output object</i>
---------------	--

Description

Prepare labels based on multiple identifiers so that they do not exceed a certain size (if they do, a simple number is used)

Usage

```
format_labels(labels, max_col = 3, max_row = 3)
```

Arguments

labels	a character vector of the labels to manage
max_col	maximum number of identifiers in a single row for a label
max_row	maximum number of rows of identifiers in a label

Value

the updated labels

Examples

```
format_labels(c("A, B", "C, D, E"))
```

gene_mutations_hist	<i>Histogram of genes' frequencies</i>
---------------------	--

Description

Create the histogram of the genes' mutational frequencies

Usage

```
gene_mutations_hist(mutmatrix, binwidth = 1)
```

Arguments

mutmatrix input dataset (mutational matrix)
binwidth binwidth parameter for the histogram (as in ggplot)

Value

the newly created histogram

Examples

```
gene_mutations_hist(example_dataset(), binwidth = 10)
```

get_no_of_children *Get number of children*

Description

Compute number of children for each node given an adj matrix

Usage

```
get_no_of_children(A, g)
```

Arguments

A Adjacency matrix of the graph g
g a graph

Value

a vector containing the number of children for each node in g

Examples

```
require(dplyr)
require(igraph)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
A <- as_adj(g)
get_no_of_children(A, g)
```

```
graph_non_transitive_subset_topology
  Default preparation of graph topology
```

Description

By default, CIMICE computes the relation between genotypes using the subset relation. For the following steps it is also important that the transitive edges are removed.

Usage

```
graph_non_transitive_subset_topology(samples, labels)
```

Arguments

samples	mutational matrix
labels	genotype labels

Value

a graph with the wanted topology

Examples

```
require(dplyr)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
graph_non_transitive_subset_topology(samples, labels)
```

```
make_dataset          Dataset line by line construction: initialization
```

Description

Initialize a dataset for "line by line" creation

Usage

```
make_dataset(...)
```

Arguments

...	gene names (do not use ' ', the input is automatically converted to strings)
-----	--

Value

a mutational matrix without samples structured as (sparse) matrix

Examples

```
make_dataset(APC,P53,KRAS)
```

```
make_generator_stub     Create a stub for a generator
```

Description

Create a generator topology directly from a dataset. The topology will follow the subset relation.

Usage

```
make_generator_stub(dataset)
```

Arguments

dataset A compacted CIMICE dataset

Value

a generator, with weight = 0 for all the edges

Examples

```
make_generator_stub(example_dataset())
```

```
make_labels             Helper function to create labels
```

Description

This function helps creating labels for nodes with different information

Usage

```
make_labels(out, mode = "samplesIDs", max_col = 3, max_row = 3)
```

Arguments

out the output object of CIMICE (es, from quick run)

mode which labels to print: samplesIDs (matching samples), sequentialIDs (just a sequential numer), geneIDs (genotype)

max_col identifiers are represented in a max_col times max_row grid (if the number of IDs exceeds, a the sequentialID number is used instead)

max_row identifiers are represented in a max_col times max_row grid (if the number of IDs exceeds, a the sequentialID number is used instead)

Value

the requested labels

Examples

```
make_labels(quick_run(example_dataset()))
```

normalizeDWNW	<i>Down weights normalization</i>
---------------	-----------------------------------

Description

Normalizes Down weights so that the sum of weights of edges exiting a node is 1

Usage

```
normalizeDWNW(g, freqs, no.of.children, A, downWeights)
```

Arguments

g	graph (a Directed Acyclic Graph)
freqs	observed genotype frequencies
no.of.children	number of children for each node
A	adjacency matrix of G
downWeights	Down weights as computed by computeDWNW

Value

a vector containing the normalized Down weights for each edge

Examples

```
require(dplyr)
require(igraph)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
# prepare adj matrix
A <- as.matrix(as_adj(g))
# pre-compute exiting edges from each node
no.of.children <- get_no_of_children(A,g)
upWeights <- computeUPW(g, freqs, no.of.children, A)
normUpWeights <- normalizeUPW(g, freqs, no.of.children, A, upWeights)
downWeights <- computeDWNW(g, freqs, no.of.children, A, normUpWeights)
normalizeUPW(g, freqs, no.of.children, A, downWeights)
```

normalizeUPW	<i>Up weights normalization</i>
--------------	---------------------------------

Description

Normalizes up weights so that the sum of weights of edges entering in a node is 1

Usage

```
normalizeUPW(g, freqs, no.of.children, A, upWeights)
```

Arguments

g	graph (a Directed Acyclic Graph)
freqs	observed genotype frequencies
no.of.children	number of children for each node
A	adjacency matrix of G
upWeights	Up weights as computed by computeUPW

Value

a vector containing the normalized Up weights for each edge

Examples

```
require(dplyr)
require(igraph)
preproc <- example_dataset() %>% dataset_preprocessing
samples <- preproc[["samples"]]
freqs <- preproc[["freqs"]]
labels <- preproc[["labels"]]
genes <- preproc[["genes"]]
g <- graph_non_transitive_subset_topology(samples, labels)
# prepare adj matrix
A <- as.matrix(as_adj(g))
# pre-compute exiting edges from each node
no.of.children <- get_no_of_children(A,g)
upWeights <- computeUPW(g, freqs, no.of.children, A)
normalizeUPW(g, freqs, no.of.children, A, upWeights)
```

perturb_dataset	<i>Perturbate a boolean matrix</i>
-----------------	------------------------------------

Description

Given a boolean matrix, randomly add False Positives (FP), False Negatives (FN) and Missing data following user defined rates. In the final matrix, missing data is represented by the value 3.

Usage

```
perturb_dataset(dataset, FP_rate = 0, FN_rate = 0, MIS_rate = 0)
```

Arguments

dataset	a matrix/sparse matrix
FP_rate	False Positive rate
FN_rate	False Negative rate
MIS_rate	Missing Data rate

Details

Note that CIMICE does not support dataset with missing data natively, so using MIS_rate != 0 will then require some pre-processing.

Value

the new, perturbed, matrix

Examples

```
require(dplyr)

example_dataset() %>%
  make_generator_stub() %>%
  set_generator_edges(
    list(
      "D", "A, D", 1 ,
      "A", "A, D", 1 ,
      "A, D", "A, C, D", 1 ,
      "A, D", "A, B, D", 1 ,
      "Clonal", "D", 1 ,
      "Clonal", "A", 1 ,
      "D", "D", 1 ,
      "A", "A", 1 ,
      "A, D", "A, D", 1 ,
      "A, C, D", "A, C, D", 1 ,
      "A, B, D", "A, B, D", 1 ,
      "Clonal", "Clonal", 1
    )) %>%
  finalize_generator %>%
  simulate_generator(3, 10) %>%
  perturb_dataset(FP_rate = 0.01, FN_rate = 0.1, MIS_rate = 0.12)
```

plot_generator

Plot a generator

Description

Simple ggraph interface to draw a generator

Usage

```
plot_generator(generator)
```

Arguments

```
generator      a generator
```

Value

a basic plot of this generator

Examples

```
require(dplyr)

example_dataset() %>%
  make_generator_stub() %>%
  set_generator_edges(
    list(
      "D", "A, D", 1 ,
      "A", "A, D", 1 ,
      "A, D", "A, C, D", 1 ,
      "A, D", "A, B, D", 1 ,
      "Clonal", "D", 1 ,
      "Clonal", "A", 1 ,
      "D", "D", 1 ,
      "A", "A", 1 ,
      "A, D", "A, D", 1 ,
      "A, C, D", "A, C, D", 1 ,
      "A, B, D", "A, B, D", 1 ,
      "Clonal", "Clonal", 1
    )) %>%
  finalize_generator %>%
  plot_generator
```

```
prepare_generator_edge_set_command
```

Prepare a command to add edge weights to a generator

Description

Prints a string in the form of the command that sets weights for all the edges of this generator.

Usage

```
prepare_generator_edge_set_command(generator, by = "labels")
```

Arguments

```
generator      a generator
by             "labels" or "samples" to use gene labels or sample labels as references for edge
              identifiers.
```

Value

NULL (the string with the function calls is printed on the stdout)

Examples

```
require(dplyr)
example_dataset() %>%
  make_generator_stub() %>%
  prepare_generator_edge_set_command()
```

prepare_labels	<i>Prepare node labels based on genotypes</i>
----------------	---

Description

Prepare node labels so that each node is labelled with a comma separated list of the altered genes representing its associated genotype.

Usage

```
prepare_labels(samples, genes)
```

Arguments

samples	input dataset (mutational matrix) as matrix
genes	list of gene names (in the columns' order)

Details

Note that after this procedure the user is expected also to run `fix_clonal_genotype` to also add the clonal genotype to the mutational matrix if it is not present.

Value

the computed edge list

Examples

```
require(dplyr)

# compact
compactDataset <- compact_dataset(example_dataset())
samples <- compactDataset$matrix

# save genes' names
genes <- colnames(compactDataset$matrix)

# keep the information on frequencies for further analysis
freqs <- compactDataset$counts/sum(compactDataset$counts)

# prepare node labels listing the mutated genes for each node
labels <- prepare_labels(samples, genes)
```

quick_run	<i>Run CIMICE defaults</i>
-----------	----------------------------

Description

This function executes CIMICE analysis on a dataset using default settings.

Usage

```
quick_run(dataset, mode = "CAPRI")
```

Arguments

dataset	a mutational matrix as a (sparse) matrix
mode	indicates the used input format. Must be either "CAPRI" or "CAPRIpop"

Value

a list object representing the graph computed by CIMICE with the structure 'list(topology = g, weights = W, labels = labels, freqs=freqs)'

Examples

```
quick_run(example_dataset())
```

read	<i>Read a "CAPRI" file</i>
------	----------------------------

Description

Read a "CAPRI" formatted file, as read_CAPRI

Usage

```
read(filepath)
```

Arguments

filepath	path to file
----------	--------------

Value

the described mutational matrix as a (sparse) matrix

Examples

```
read(system.file("extdata", "example.CAPRI", package = "CIMICE", mustWork = TRUE))
```

read_CAPRI	<i>Read a "CAPRI" file</i>
------------	----------------------------

Description

Read a "CAPRI" formatted file from the file system

Usage

```
read_CAPRI(filepath)
```

Arguments

filepath path to file

Value

the described mutational matrix as a (sparse) matrix

Examples

```
#            "pathToDataset/myDataset.CAPRI"  
read_CAPRI(system.file("extdata", "example.CAPRI", package = "CIMICE", mustWork = TRUE))
```

read_CAPRIpop	<i>Read a "CAPRIpop" file</i>
---------------	-------------------------------

Description

Read a "CAPRIpop" formatted file from the file system

Usage

```
read_CAPRIpop(filepath)
```

Arguments

filepath path to file

Value

a list containing the described mutational matrix as a (sparse) matrix and a list of the frequency of the genotypes

Examples

```
#            "pathToDataset/myDataset.CAPRI"  
read_CAPRI(system.file("extdata", "example.CAPRIpop", package = "CIMICE", mustWork = TRUE))
```

read_CAPRIpop_string *Read "CAPRIpop" file from a string*

Description

Read a "CAPRIpop" formatted file, from a text string

Usage

```
read_CAPRIpop_string(txt)
```

Arguments

txt string in valid "CAPRIpop" format

Value

the described mutational matrix as a (sparse) matrix

Examples

```
read_CAPRIpop_string("
s\\g A B C D freqs
S1 0 0 0 1 0.1
S2 1 0 0 0 0.1
S3 1 0 0 0 0.2
S4 1 0 0 1 0.3
S5 1 1 0 1 0.05
S6 1 1 0 1 0.1
S7 1 0 1 1 0.05
S8 1 1 0 1 0.01
")
```

read_CAPRI_string *Read "CAPRI" file from a string*

Description

Read a "CAPRI" formatted file, from a text string

Usage

```
read_CAPRI_string(txt)
```

Arguments

txt string in valid "CAPRI" format

Value

the described mutational matrix as a (sparse) matrix

Examples

```
read_CAPRI_string("
s\\g A B C D
S1 0 0 0 1
S2 1 0 0 0
S3 1 0 0 0
S4 1 0 0 1
S5 1 1 0 1
S6 1 1 0 1
S7 1 0 1 1
S8 1 1 0 1
")
```

read_MAF

Create mutational matrix from MAF file

Description

Read a MAF (Mutation Annotation Format) file and create a Mutational Matrix combining gene and sample IDs.

Usage

```
read_MAF(path, ...)
```

Arguments

path	path to MAF file
...	other maftools::mutCountMatrix arguments

Value

the mutational (sparse) matrix associated to the MAF file

Examples

```
read_MAF(system.file("extdata", "paac_jhu_2014_500.maf", package = "CIMICE", mustWork = TRUE))
```

read_matrix	<i>Read dataset from an R matrix</i>
-------------	--------------------------------------

Description

also converts that matrix to a sparse matrix

Usage

```
read_matrix(mat)
```

Arguments

mat a boolean mutational matrix

Value

the sparse mutational matrix to be used as CIMICE's input

Examples

```
m <- matrix(c(0,0,1,1,0,1,1,1,1), ncol=3)
colnames(m) <- c("A", "B", "C")
rownames(m) <- c("S1", "S2", "S3")
read_matrix(m)
```

remove_transitive_edges	<i>Remove transitive edges from an edgelist</i>
-------------------------	---

Description

Remove transitive edges from an edgelist. This procedure is temporary to cover a bug in 'relations' package.

Usage

```
remove_transitive_edges(E)
```

Arguments

E edge list, built from "build_topology_subset"

Value

a new edgelist without transitive edges (as a N*2 matrix)

Examples

```
l <- list(c(1,2),c(2,3), c(1,3))
remove_transitive_edges(l)
```

sample_mutations_hist *Histogram of samples' frequencies*

Description

Create the histogram of the samples' mutational frequencies

Usage

```
sample_mutations_hist(mutmatrix, binwidth = 1)
```

Arguments

mutmatrix	input dataset (mutational matrix)
binwidth	binwidth parameter for the histogram (as in ggplot)

Value

the newly created histogram

Examples

```
sample_mutations_hist(example_dataset(), binwidth = 10)
```

select_genes_on_mutations
Filter dataset by genes' mutation count

Description

Dataset filtering on genes, based on their mutation count

Usage

```
select_genes_on_mutations(mutmatrix, n, desc = TRUE)
```

Arguments

mutmatrix	input dataset (mutational matrix) to be reduced
n	number of genes to be kept
desc	TRUE: select the n least mutated genes, FALSE: select the n most mutated genes

Value

the modified dataset (mutational matrix)

Examples

```
# keep information on the 100 most mutated genes
select_genes_on_mutations(example_dataset(), 5)
# keep information on the 100 least mutated genes
select_genes_on_mutations(example_dataset(), 5, desc = FALSE)
```

```
select_samples_on_mutations
      Filter dataset by samples' mutation count
```

Description

Dataset filtering on samples, based on their mutation count

Usage

```
select_samples_on_mutations(mutmatrix, n, desc = TRUE)
```

Arguments

mutmatrix	input dataset (mutational matrix) to be reduced
n	number of samples to be kept
desc	T: select the n least mutated samples, F: select the n most mutated samples

Value

the modified dataset (mutational matrix)

Examples

```
require(dplyr)
# keep information on the 5 most mutated samples
select_samples_on_mutations(example_dataset(), 5)
# keep information on the 5 least mutated samples
select_samples_on_mutations(example_dataset(), 5, desc = FALSE)
# combine selections
select_samples_on_mutations(example_dataset(), 5, desc = FALSE) %>%
  select_genes_on_mutations(5)
```

set_generator_edges *Add edge weights to a generator*

Description

Add edge weights to a generator

Usage

```
set_generator_edges(generator, f_t_w_list, by = "labels")
```

Arguments

generator	a generator
f_t_w_list	a list of triplets (from, to, list), the triplets must not be nested in the list. For example list("A","B",0.3, "B", "C", 0.2) is a valid input.
by	"labels" or "samples" to use gene labels or sample labels as references for edge identifiers.

Value

the generator with the modified edges (invalid edges are ignored)

Examples

```
require(dplyr)

example_dataset() %>%
  make_generator_stub() %>%
  set_generator_edges(
    list(
      "D", "A, D", 1 ,
      "A", "A, D", 1 ,
      "A, D", "A, C, D", 1 ,
      "A, D", "A, B, D", 1 ,
      "Clonal", "D", 1 ,
      "Clonal", "A", 1 ,
      "D", "D", 1 ,
      "A", "A", 1 ,
      "A, D", "A, D", 1 ,
      "A, C, D", "A, C, D", 1 ,
      "A, B, D", "A, B, D", 1 ,
      "Clonal", "Clonal", 1
    )
  )
```

simulate_generator *Create datasets from generators*

Description

Simulate the DTMC associated to the generator to create a dataset that reflects the genotypes of 'times' cells, sampled after 'time_ticks' passages.

Usage

```
simulate_generator(
  generator,
  time_ticks,
  times,
  starting_label = "Clonal",
  by = "labels",
  mode = "full"
)
```

Arguments

generator	a generator
time_ticks	number of steps (updates) of the DTMC associated to the generato
times	number of sumlated cells
starting_label	node from which to start the simulation
by	"labels" or "samples" to use gene labels or sample labels as references to identify the 'starting_label's node
mode	"full" to generate a matrix with 'times' genotypes, "compacted" to *efficiently* create an already compacted dataset (a dataset showing the genotypes and their respective frequencies)

Value

the simulated dataset

Examples

```
require(dplyr)

example_dataset() %>%
  make_generator_stub() %>%
  set_generator_edges(
    list(
      "D", "A, D", 1 ,
      "A", "A, D", 1 ,
      "A, D", "A, C, D", 1 ,
      "A, D", "A, B, D", 1 ,
      "Clonal", "D", 1 ,
      "Clonal", "A", 1 ,
      "D", "D", 1 ,
      "A", "A", 1 ,
    )
  )
```

```

    "A, D", "A, D", 1 ,
    "A, C, D", "A, C, D", 1 ,
    "A, B, D", "A, B, D", 1 ,
    "Clonal", "Clonal", 1
  )) %>%
  finalize_generator %>%
  simulate_generator(3, 10)

```

to_dot

Dot graph output

Description

Represents this graph in dot format (a textual output format)

Usage

```
to_dot(out, ...)
```

Arguments

out the output object of CIMICE (es, from quick run)
 ... other arguments for format_labels

Value

a string representing the graph in dot format

Examples

```
to_dot(quick_run(example_dataset()))
```

update_df

Dataset line by line construction: add a sample

Description

Add a sample (a row) to an existing dataset. This procedure is meant to be used with the "

Usage

```
update_df(mutmatrix, sampleName, ...)
```

Arguments

mutmatrix an existing (sparse) matrix (mutational matrix)
 sampleName the row (sample) name
 ... sample's genotype (0/1 numbers)

Value

the modified (sparse) matrix (mutational matrix)

Examples

```
require(dplyr)
make_dataset(APC,P53,KRAS) %>%
  update_df("S1", 1, 0, 1) %>%
  update_df("S2", 1, 1, 1)
```

Index

annotate_mutational_matrix, 3

binary_radix_sort, 4
build_subset_graph, 4
build_topology_subset, 5

chunk_reader, 6
CIMICE, 6
compact_dataset, 7
compute_weights_default, 10
computeDWNW, 7
computeDWNW_aux, 8
computeUPW, 9
computeUPW_aux, 10
corrplot_from_mutational_matrix, 11
corrplot_genes, 11
corrplot_samples, 12

dataset_preprocessing, 12
dataset_preprocessing_population, 13
draw_ggraph, 14
draw_networkD3, 14
draw_visNetwork, 15

example_dataset, 15
example_dataset_withFreqs, 16

finalize_generator, 16
fix_clonal_genotype, 17
format_labels, 18

gene_mutations_hist, 18
get_no_of_children, 19
graph_non_transitive_subset_topology,
20

make_dataset, 20
make_generator_stub, 21
make_labels, 21

normalizeDWNW, 22
normalizeUPW, 23

perturb_dataset, 23
plot_generator, 24

prepare_generator_edge_set_command, 25
prepare_labels, 26

quick_run, 27

read, 27
read_CAPRI, 28
read_CAPRI_string, 29
read_CAPRIpop, 28
read_CAPRIpop_string, 29
read_MAF, 30
read_matrix, 31
remove_transitive_edges, 31

sample_mutations_hist, 32
select_genes_on_mutations, 32
select_samples_on_mutations, 33
set_generator_edges, 34
simulate_generator, 35

to_dot, 36

update_df, 36