

Package ‘netboost’

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Type Package

Title Network Analysis Supported by Boosting

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Description Boosting supported network analysis for high-dimensional omics applications.
This package comes bundled with the MC-UPGMA clustering package by Yaniv Loewenstein.

Depends R (>= 4.0.0)

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Clustering, DimensionReduction, BiomedicalInformatics,
Epigenetics, Metabolomics, Transcriptomics

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R.utils

LinkingTo Rcpp, RcppParallel

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BugReports <mailto:pascal.schlosser@uniklinik-freiburg.de>

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Contents

.onAttach	3
calculate_adjacency	3
cpp_dist_tom	4
cpp_filter_base	5
cpp_filter_end	5
cpp_filter_step	6
cpp_tree_search	6
cut_dendro	7
cut_trees	8
mcupgma_exec	9
nb_clust	10
nb_consensus	11
nb_consensus_transfer	13
nb_dist	14
nb_filter	15
nb_mcupgma	16
nb_moduleEigengenes	17
nb_plot_dendro	20
nb_summary	21
nb_transfer	21
netboost	22
netboostMCUPGMAPath	24
netboostPackagePath	25
netboostTmpCleanup	25
netboostTmpPath	26
tcga_aml_meth_rna_chr18	26
tree_dendro	27
tree_search	27

Index	29
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<code>.onAttach</code>	<i>Package startup: used to fetch installation path of the own package, as required for executing binary programs delivered with it.</i>
------------------------	--

Description

Package startup: used to fetch installation path of the own package, as required for executing binary programs delivered with it.

Usage

```
.onAttach(libname, pkgname)
```

Arguments

<code>libname</code>	Path to R installation (base package dir)
<code>pkgname</code>	Package name (should be "netboost")

Value

none

Examples

```
## Not run: nb_example()
```

<code>calculate_adjacency</code>	<i>Calculate network adjacencies for filter</i>
----------------------------------	---

Description

Calculate network adjacencies for filter

Usage

```
calculate_adjacency(  
  datan,  
  filter,  
  soft_power = 2,  
  method = c("pearson", "kendall", "spearman"),  
  cores = getOption("mc.cores", 2L)  
)
```

Arguments

datan	Data frame were rows correspond to samples and columns to features.
filter	Filter-Matrix as generated by the nb_filter function.
soft_power	Integer. Exponent of the transformation. Set automatically based on the scale free topology criterion if unspecified.
method	A character string specifying the method to be used for correlation coefficients.
cores	Deprecated. Kept for backward compatibility but unused.

Value

Numeric vector with adjacencies for the filter

cpp_dist_tom	<i>Function to calculate distance</i>
--------------	---------------------------------------

Description

Function to calculate distance

Usage

```
cpp_dist_tom(filter, adjacency)
```

Arguments

filter	Filter matrix
adjacency	Vector

Details

Steps: 1. - Sequential preparation of index and partner caches per value in filter 2. - Parallel calculation of the distances with cached vectors

Value

numeric vector

cpp_filter_base	<i>Initialise boosting with chosen accelerator hardware (x86, AVX, FMA)</i>
-----------------	---

Description

Initialise boosting with chosen accelerator hardware (x86, AVX, FMA)

Usage

```
cpp_filter_base(data, stepno = 20L, mode_ = 2L)
```

Arguments

data	Matrix
stepno	Amount of steps
mode_	Accelerator mode (0: x86, 1: FMA, 2: AVX)

Value

none

cpp_filter_end	<i>Boosting cleanup (required to free memory)</i>
----------------	---

Description

Boosting cleanup (required to free memory)

Usage

```
cpp_filter_end()
```

Value

none

cpp_filter_step	<i>Single boosting step</i>
-----------------	-----------------------------

Description

Single boosting step

Usage

```
cpp_filter_step(col_y)
```

Arguments

col_y	Row in data matrix
-------	--------------------

Details

Must be initialised before using [cpp_filter_base](#)

Value

integer vector

cpp_tree_search	<i>Tree search.</i>
-----------------	---------------------

Description

Constraint: IDs $0 \leq x$ (Integer)

Usage

```
cpp_tree_search(netboost_forest)
```

Arguments

netboost_forest	Input-matrix (4 columns, ids in column 0,1,3)
-----------------	---

Value

List

cut_dendro	<i>Module detection for an individual tree</i>
------------	--

Description

Module detection for an individual tree

Usage

```
cut_dendro(
  tree_dendro,
  min_cluster_size = 2L,
  datan,
  ME_diss_thres,
  name_of_tree = "",
  qc_plot = TRUE,
  n_pc = 1,
  robust_PCs = FALSE,
  nb_min_varExpl = 0.5,
  method = c("pearson", "kendall", "spearman")
)
```

Arguments

tree_dendro	List of tree specific objects including dendrogram, tree data and features originating from the tree_dendro function.
min_cluster_size	Integer. The minimum number of features in one module.
datan	Data frame where rows correspond to samples and columns to features.
ME_diss_thres	Numeric. Module Eigengene Dissimilarity Threshold for merging close modules.
name_of_tree	String. Annotating plots and messages.
qc_plot	Logical. Should plots be created?
n_pc	Number of principal components and variance explained entries to be calculated. The number of returned variance explained entries is currently 'min(n_pc,10)'. If given 'n_pc' is greater than 10, a warning is issued.
robust_PCs	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
nb_min_varExpl	Minimum proportion of variance explained for returned module eigengenes. The number of PCs is capped at n_pc.
method	A character string specifying the method to be used for correlation coefficients.

Value

List

cut_trees

Module detection for the results from a nb_mcupgma call

Description

Module detection for the results from a nb_mcupgma call

Usage

```
cut_trees(
  trees,
  datan,
  forest,
  min_cluster_size = 2L,
  ME_diss_thres,
  qc_plot = TRUE,
  n_pc = 1,
  robust_PCs = FALSE,
  nb_min_varExpl = 0.5,
  method = c("pearson", "kendall", "spearman")
)
```

Arguments

trees	List of trees, where one tree is a list of ids and rows
datan	Data frame where rows correspond to samples and columns to features.
forest	Raw dendrogram-matrix as generated by the nb_mcupgma function.
min_cluster_size	Integer. The minimum number of features in one module.
ME_diss_thres	Numeric. Module Eigengene Dissimilarity Threshold for merging close modules.
qc_plot	Logical. Should plots be created?
n_pc	Number of principal components and variance explained entries to be calculated. The number of returned variance explained entries is currently 'min(n_pc,10)'. If given 'n_pc' is greater than 10, a warning is issued.
robust_PCs	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
nb_min_varExpl	Minimum proportion of variance explained for returned module eigengenes. The number of PCs is capped at n_pc.
method	A character string specifying the method to be used for correlation coefficients.

Value

List

Examples

```

data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
cores=cores,mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
max_singleton = dim(tcga_aml_meth_rna_chr18)[2]
forest <- nb_mcupgma(filter=filter, dist=dist, max_singleton=max_singleton,
cores=cores)
trees <- tree_search(forest)
results <- cut_trees(trees=trees,datan=datan, forest=forest,
min_cluster_size=10L, ME_diss_thres=0.25, qc_plot=TRUE)

```

mcupgma_exec

*Execute a program/script from the installed MCUPGMA suite.***Description**

Execute a program/script from the installed MCUPGMA suite.

Usage

```
mcupgma_exec(exec = NULL, ..., console = TRUE)
```

Arguments

exec	Name of the file of the executable.
...	Arguments passed to mcupgma executable in order required by program
console	Logical. Print output to R console or fetch for return to caller

Value

console=TRUE: exit code (0: no error). console=FALSE: STDOUT/STDERR output

Examples

```
mcupgma_exec(exec="cluster.pl", "--help")
```

nb_clust	<i>Netboost clustering step</i>
----------	---------------------------------

Description

Netboost clustering step

Usage

```
nb_clust(
  filter,
  dist,
  datan,
  max_singleton = dim(datan)[2],
  min_cluster_size = 2L,
  ME_diss_thres = 0.25,
  cores = getOption("mc.cores", 2L),
  qc_plot = TRUE,
  n_pc = 1,
  robust_PCs = FALSE,
  nb_min_varExpl = 0.5,
  method = c("pearson", "kendall", "spearman")
)
```

Arguments

filter	Filter-Matrix as generated by the nb_filter function.
dist	Distance-Matrix as generated by the nb_dist function.
datan	Data frame where rows correspond to samples and columns to features.
max_singleton	Integer. The maximal singleton in the clustering. Usually equals the number of features.
min_cluster_size	Integer. The minimum number of features in one module.
ME_diss_thres	Numeric. Module Eigengene Dissimilarity Threshold for merging close modules.
cores	Integer. Amount of CPU cores used (<=1 : sequential)
qc_plot	Logical. Create plot.
n_pc	Number of principal components and variance explained entries to be calculated. The number of returned variance explained entries is currently 'min(n_pc,10)'. If given 'n_pc' is greater than 10, a warning is issued.
robust_PCs	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
nb_min_varExpl	Minimum proportion of variance explained for returned module eigengenes. The number of PCs is capped at n_pc.
method	A character string specifying the method to be used for correlation coefficients.

Value

List

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
cores=cores,mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
max_singleton = dim(tcga_aml_meth_rna_chr18)[2]
pdf("test.pdf",width=30)
sum_res <- nb_clust(filter=filter, dist=dist, datan=datan,
max_singleton=max_singleton, min_cluster_size=10L, ME_diss_thres=0.25,
cores=cores, qc_plot=TRUE, n_pc=2L, nb_min_varExpl=0.5)
dev.off()
```

nb_consensus

*Multi-dataset consensus network analysis***Description**

Performs a 10-step consensus network analysis across multiple datasets: (1-2) Filtered adjacency per dataset, (3-4) Direction filter + consensus, (5) Soft thresholding, (6) TOM, (7) Hierarchical clustering, (8) Module eigengenes, (9) Cross-dataset module merging, (10) Module retention.

Usage

```
nb_consensus(
  datan_list,
  filter_method = c("spearman", "pearson", "kendall", "boosting", "skip"),
  soft_power = NULL,
  consensus_method = c("min", "max"),
  network_type = "unsigned",
  filter_dir = TRUE,
  min_cluster_size = 15L,
  ME_diss_thres = 0.25,
  merge_criterion = c("all", "any"),
  module_retention = c("any", "all"),
  deepSplit = 2,
  method = c("pearson", "kendall", "spearman"),
  scale = TRUE,
  stepno = 20L,
  n_pc = 1,
  robust_PCs = FALSE,
  nb_min_varExpl = 0.5,
```

```

cores = 1L,
pdf_report = NULL,
mask_cache_dir = NULL,
verbose = TRUE
)

```

Arguments

<code>datan_list</code>	List of data frames or matrices (rows = samples, cols = features). All datasets must share the same column names.
<code>filter_method</code>	Filtering method: "spearman", "pearson", "kendall", "boosting", or "skip".
<code>soft_power</code>	Numeric soft-thresholding power, or NULL for none.
<code>consensus_method</code>	How to combine adjacencies across datasets: "min" or "max".
<code>network_type</code>	Network type (currently only "unsigned").
<code>filter_dir</code>	Logical. Apply direction filter (require consistent correlation sign across datasets)?
<code>min_cluster_size</code>	Integer. Minimum module size.
<code>ME_diss_thres</code>	Numeric (0, 1). Module eigengene dissimilarity threshold for merging.
<code>merge_criterion</code>	How to evaluate cross-dataset ME correlation for merging: "all" (pmin) or "any" (pmax).
<code>module_retention</code>	Criterion for retaining modules: "any" or "all".
<code>deepSplit</code>	Integer passed to cutreeDynamic .
<code>method</code>	Correlation method for adjacency calculation.
<code>scale</code>	Logical. Scale and center data before PCA?
<code>stepno</code>	Integer. Number of boosting steps for nb_filter .
<code>n_pc</code>	Integer. Number of principal components.
<code>robust_PCs</code>	Logical. Use rank-based (Spearman) PCA?
<code>nb_min_varExpl</code>	Numeric. Minimum variance explained for module eigengenes.
<code>cores</code>	Integer. Number of CPU cores.
<code>pdf_report</code>	Path for PDF report, or NULL to skip.
<code>mask_cache_dir</code>	Directory for caching filter masks, or NULL for <code>tempdir()</code> .
<code>verbose</code>	Logical. Print progress messages?

Value

A list with components:

names Feature names.

colors Final module color assignments (WGCNA color labels).

MEs Module eigengenes for the first dataset.

var_explained Variance explained per dataset.
rotation Mean PC rotation matrix across datasets.
filter Per-dataset filter edges.
colors_initial Pre-merge module assignments.
colors_merged Post-merge, pre-retention assignments.
modules_retained Retained module color labels.
geneTree Hierarchical clustering dendrogram.
ME_all_datasets Module eigengenes per dataset.
ME_cor_per_dataset ME correlation matrices per dataset.
ME_tree ME dendrogram.
pc1_weights PC1 weight comparison across datasets.
filter_stats Edge filtering statistics.
params Analysis parameters.
n_datasets Number of datasets.
n_features Number of features.
n_modules_initial Modules before merging.
n_modules_merged Modules after merging.
n_modules_final Modules after retention.

Examples

```

data('tcga_aml_meth_rna_chr18', package='netboost')
# Create two mock datasets from the example data
d1 <- tcga_aml_meth_rna_chr18[1:40, 1:50]
d2 <- tcga_aml_meth_rna_chr18[41:80, 1:50]
res <- nb_consensus(datan_list = list(d1, d2),
  filter_method = "spearman", stepno = 20L,
  min_cluster_size = 5L, ME_diss_thres = 0.25,
  verbose = FALSE)

```

nb_consensus_transfer *Transfer consensus module assignments to new data*

Description

Projects consensus module eigengenes onto new samples using the rotation matrix from [nb_consensus](#).

Usage

```

nb_consensus_transfer(
  nb_consensus_result,
  new_data,
  scale = FALSE,
  robust PCs = FALSE
)

```

Arguments

nb_consensus_result Result from [nb_consensus](#).

new_data Data frame or matrix (rows = samples, cols = features).

scale Logical. Scale and center new_data before projection?

robust_PCs Logical. Use rank-based PCA (Spearman)?

Value

Data frame of projected module eigengenes (samples x modules).

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
d1 <- tcga_aml_meth_rna_chr18[1:40, 1:50]
d2 <- tcga_aml_meth_rna_chr18[41:80, 1:50]
res <- nb_consensus(datan_list = list(d1, d2),
  filter_method = "spearman", stepno = 20L,
  min_cluster_size = 5L, ME_diss_thres = 0.25,
  verbose = FALSE)
new_dat <- tcga_aml_meth_rna_chr18[1:10, 1:50]
MEs_new <- nb_consensus_transfer(res, new_dat)
```

nb_dist	<i>Calculate distance (external wrapper for internal C++ function) Parallelisation inside C++ program with RcppParallel.</i>
---------	--

Description

Calculate distance (external wrapper for internal C++ function) Parallelisation inside C++ program with RcppParallel.

Usage

```
nb_dist(
  filter,
  datan,
  soft_power = 2,
  cores = getOption("mc.cores", 2L),
  verbose = getOption("verbose"),
  method = c("pearson", "kendall", "spearman")
)
```

Arguments

filter	Filter-Matrix as generated by the nb_filter function.
datan	Data frame where rows correspond to samples and columns to features.
soft_power	Integer. Exponent of the transformation. Set automatically based on the scale free topology criterion if unspecified.
cores	Integer. Amount of CPU cores used (≤ 1 : sequential).
verbose	Additional diagnostic messages.
method	A character string specifying the method to be used for correlation coefficients.

Value

Vector with distances.

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
cores=cores, mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
summary(dist)
```

nb_filter	<i>Boosting via C++ function. Parallelisation by R-package parallel with forking (overhead of this method does not fall into account as single steps are ~10s).</i>
-----------	---

Description

Parallelisation via multicore (via 'parallel'-package). So *nix only atm.

Usage

```
nb_filter(
  datan,
  stepno = 20L,
  until = 0L,
  progress = 1000L,
  filter_method = c("spearman", "skip", "kendall", "boosting", "pearson"),
  cores = getOption("mc.cores", 2L),
  mode = 2L,
  verbose = getOption("verbose")
)
```

Arguments

<code>datan</code>	Data frame where rows correspond to samples and columns to features.
<code>stepno</code>	Integer amount of boosting steps
<code>until</code>	Stop at index/column (if 0: iterate through all columns)
<code>progress</code>	Integer. If > 0, print progress after every X steps (mind: parallel!)
<code>filter_method</code>	The following filtering methods are supported: "boosting" (non-zero coefficients in likelihood based boosting), "skip" (no filter), "kendall" (stats::cor.test), "spearman" (stats::cor.test), "pearson" (stats::cor.test)
<code>cores</code>	Integer. Amount of CPU cores used (<=1 : sequential)
<code>mode</code>	Integer. Mode (0: x86, 1: FMA, 2: AVX). Features are only available if compiled accordingly and available on the hardware.
<code>verbose</code>	Additional diagnostic messages.

Value

matrix n times 2 matrix with the indices of the n unique entries of the filter

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
cores=cores, mode=2L)
head(filter)
nrow(filter)/(ncol(datan)*(ncol(datan)-1)/2) # proportion of potential undirected edges
```

nb_mcupgma

Calculate dendrogram for a sparse distance matrix (external wrapper MC-UPGMA clustering package Loewenstein et al.

Description

Calculate dendrogram for a sparse distance matrix (external wrapper MC-UPGMA clustering package Loewenstein et al.

Usage

```
nb_mcupgma(
  filter,
  dist,
  max_singleton,
  cores = getOption("mc.cores", 2L),
  verbose = getOption("verbose")
)
```

Arguments

filter	Filter-Matrix as generated by the nb_filter function.
dist	Distance-Matrix as generated by the nb_dist function.
max_singleton	Integer The maximal singleton in the clustering. Usually equals the number of features.
cores	Integer Amount of CPU cores used (<=1 : sequential)
verbose	Logical Additional diagnostic messages.

Value

Raw dendrogram to be processed by tree_search and tree_dendro.

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18,
  center=TRUE, scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L,
  progress=1000L, cores=cores, mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
max_singleton = dim(tcga_aml_meth_rna_chr18)[2]
forest <- nb_mcupgma(filter=filter, dist=dist,
  max_singleton=max_singleton, cores=cores)
head(forest)
```

nb_moduleEigengenes *Netboost module aggregate extraction.*

Description

This is a modification of WGCNA::moduleEigengenes() (version WGCNA_1.66) to include more than the first principal component. For details see WGCNA::moduleEigengenes().

Usage

```
nb_moduleEigengenes(
  expr,
  colors,
  n_pc = 1,
  align = "along average",
  exclude_grey = FALSE,
  grey = if (is.numeric(colors)) 0 else "grey",
  subHubs = TRUE,
  robust = FALSE,
  trapErrors = FALSE,
```

```

    return_valid_only = trapErrors,
    soft_power = 6,
    scale = TRUE,
    verbose = 0,
    indent = 0,
    nb_min_varExpl = 0.5
)

```

Arguments

<code>expr</code>	Expression data for a single set in the form of a data frame where rows are samples and columns are genes (probes).
<code>colors</code>	A vector of the same length as the number of probes in 'expr', giving module color for all probes (genes). Color "grey" is reserved for unassigned genes. Expression
<code>n_pc</code>	Number of principal components and variance explained entries to be calculated. The number of returned variance explained entries is currently 'min(n_pc,10)'. If given 'n_pc' is greater than 10, a warning is issued.
<code>align</code>	Controls whether eigengenes, whose orientation is undetermined, should be aligned with average expression ('align = 'along average', the default) or left as they are ('align = '). Any other value will trigger an error.
<code>exclude_grey</code>	Should the improper module consisting of 'grey' genes be excluded from the eigengenes?
<code>grey</code>	Value of 'colors' designating the improper module. Note that if 'colors' is a factor of numbers, the default value will be incorrect.
<code>subHubs</code>	Controls whether hub genes should be substituted for missing eigengenes. If 'TRUE', each missing eigengene (i.e., eigengene whose calculation failed and the error was trapped) will be replaced by a weighted average of the most connected hub genes in the corresponding module. If this calculation fails, or if 'subHubs==FALSE', the value of 'trapErrors' will determine whether the offending module will be removed or whether the function will issue an error and stop.
<code>robust</code>	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
<code>trapErrors</code>	Controls handling of errors from that may arise when there are too many 'NA' entries in expression data. If 'TRUE', errors from calling these functions will be trapped without abnormal exit. If 'FALSE', errors will cause the function to stop. Note, however, that 'subHubs' takes precedence in the sense that if 'subHubs==TRUE' and 'trapErrors==FALSE', an error will be issued only if both the principal component and the hubgene calculations have failed.
<code>return_valid_only</code>	logical; controls whether the returned data frame of module eigengenes contains columns corresponding only to modules whose eigengenes or hub genes could be calculated correctly ('TRUE'), or whether the data frame should have columns for each of the input color labels ('FALSE').

soft_power	The power used in soft-thresholding the adjacency matrix. Only used when the hubgene approximation is necessary because the principal component calculation failed. It must be non-negative. The default value should only be changed if there is a clear indication that it leads to incorrect results.
scale	logical; can be used to turn off scaling of the expression data before calculating the singular value decomposition. The scaling should only be turned off if the data has been scaled previously, in which case the function can run a bit faster. Note however that the function first imputes, then scales the expression data in each module. If the expression contain missing data, scaling outside of the function and letting the function impute missing data may lead to slightly different results than if the data is scaled within the function.
verbose	Controls verbosity of printed progress messages. 0 means silent, up to (about) 5 the verbosity gradually increases.
indent	A single non-negative integer controlling indentation of printed messages. 0 means no indentation, each unit above that adds two spaces.
nb_min_varExpl	Minimum proportion of variance explained for returned module eigengenes. Is capped at n_pc.

Value

eigengenes Module eigengenes in a dataframe, with each column corresponding to one eigengene. The columns are named by the corresponding color with an "ME" prepended, e.g., 'MEturquoise' etc. If 'return_valid_only==FALSE', module eigengenes whose calculation failed have all components set to 'NA'.

averageExpr If 'align == 'along average'', a dataframe containing average normalized expression in each module. The columns are named by the corresponding color with an "AE" prepended, e.g., 'Aeturquoise' etc.

var_explained A dataframe in which each column corresponds to a module, with the component 'var_explained[PC, module]' giving the variance of module 'module' explained by the principal component no. 'PC'. The calculation is exact irrespective of the number of computed principal components. At most 10 variance explained values are recorded in this dataframe.

n_pc A copy of the input 'n_pc'.

validMEs A boolean vector. Each component (corresponding to the columns in 'data') is 'TRUE' if the corresponding eigengene is valid, and 'FALSE' if it is invalid. Valid eigengenes include both principal components and their hubgene approximations. When 'return_valid_only==FALSE', by definition all returned eigengenes are valid and the entries of 'validMEs' are all 'TRUE'.

validColors A copy of the input colors with entries corresponding to invalid modules set to 'grey' if given, otherwise 0 if 'colors' is numeric and 'grey' otherwise.

allOK Boolean flag signalling whether all eigengenes have been calculated correctly, either as principal components or as the hubgene average approximation.

allPC Boolean flag signalling whether all returned eigengenes are principal components.

isPC Boolean vector. Each component (corresponding to the columns in 'eigengenes') is 'TRUE' if the corresponding eigengene is the first principal component and 'FALSE' if it is the hubgene approximation or is invalid.

isHub Boolean vector. Each component (corresponding to the columns in 'eigengenes') is 'TRUE' if the corresponding eigengene is the hubgene approximation and 'FALSE' if it is the first principal component or is invalid.

validAEs Boolean vector. Each component (corresponding to the columns in 'eigengenes') is 'TRUE' if the corresponding module average expression is valid.

allAEOK Boolean flag signalling whether all returned module average expressions contain valid data. Note that 'return_valid_only==TRUE' does not imply 'allAEOK==TRUE': some invalid average expressions may be returned if their corresponding eigengenes have been calculated correctly.

nb_plot_dendro	<i>Plot dendrogram from Netboost output.</i>
----------------	--

Description

Plot dendrogram from Netboost output.

Usage

```
nb_plot_dendro(
  nb_summary = NULL,
  labels = FALSE,
  main = "",
  colorsrandom = FALSE
)
```

Arguments

nb_summary	Netboost results as generated by the nb_summary function.
labels	Boolean flag whether labels should be attached to the leafs.
main	Plot title.
colorsrandom	Boolean flag whether module colors should be shuffled.

Value

invisible null

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
results <- netboost(datan = tcga_aml_meth_rna_chr18, stepno = 20L,
  soft_power = 3L, min_cluster_size = 10L, n_pc = 2, scale=TRUE,
  ME_diss_thres = 0.25, qc_plot = FALSE)
set.seed(1234) # reproducible but shuffled color-module matching
nb_plot_dendro(nb_summary = results, labels = FALSE, main = 'Test',
  colorsrandom = TRUE)
```

nb_summary

Summarize results from a forest. Plot trees together.

Description

Summarize results from a forest. Plot trees together.

Usage

```
nb_summary(clust_res)
```

Arguments

clust_res Clustering results from cut_trees call.

Value

List

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
cores=cores,mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
max_singleton = dim(tcga_aml_meth_rna_chr18)[2]
forest <- nb_mcupgma(filter=filter,dist=dist,max_singleton=max_singleton,
cores=cores)
trees <- tree_search(forest)
results <- cut_trees(trees=trees,datan=datan, forest=forest,
min_cluster_size=10L, ME_diss_thres=0.25, qc_plot=FALSE)
sum_res <- nb_summary(clust_res=results)
```

nb_transfer

Transfer of Netboost clustering to new data.

Description

Transfer of Netboost clustering to new data.

Usage

```
nb_transfer(
  nb_summary = NULL,
  new_data = NULL,
  scale = FALSE,
  robust_PCs = FALSE,
  only_module_membership = FALSE
)
```

Arguments

<code>nb_summary</code>	Netboost results as generated by the <code>nb_summary</code> function.
<code>new_data</code>	Data frame where rows correspond to samples and columns to features.
<code>scale</code>	Logical. Should data be scaled and centered?
<code>robust_PCs</code>	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
<code>only_module_membership</code>	Logical. Should only module memberships be transferred and PCs be newly computed?

Value

List

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
results <- netboost(datan = tcga_aml_meth_rna_chr18, stepno = 20L,
  soft_power = 3L, min_cluster_size = 10L, n_pc = 2, scale=TRUE,
  ME_diss_thres = 0.25, qc_plot=FALSE)
ME_transfer <- nb_transfer(nb_summary = results,
  new_data = tcga_aml_meth_rna_chr18,
  scale = TRUE)
all(round(results[["MEs"]], 12) == round(ME_transfer, 12))
```

netboost

Netboost clustering.

Description

The Netboost clustering is performed in three subsequent steps. First, a filter of important edges in the network is calculated. Next, pairwise distances are calculated. Last, clustering is performed. For details see Schlosser et al. doi...

Usage

```

netboost(
  datan = NULL,
  stepno = 20L,
  filter_method = c("boosting", "skip", "kendall", "spearman", "pearson"),
  until = 0L,
  progress = 1000L,
  mode = 2L,
  soft_power = NULL,
  max_singleton = ncol(datan),
  qc_plot = TRUE,
  min_cluster_size = 2L,
  ME_diss_thres = 0.25,
  n_pc = 1,
  robust PCs = FALSE,
  nb_min_varExpl = 0.5,
  cores = as.integer(getOption("mc.cores", 2)),
  scale = TRUE,
  method = c("pearson", "kendall", "spearman"),
  verbose = getOption("verbose")
)

```

Arguments

<code>datan</code>	Data frame where rows correspond to samples and columns to features.
<code>stepno</code>	Integer amount of boosting steps applied in the filtering step
<code>filter_method</code>	The following filtering methods are supported: "boosting" (non-zero coefficients in likelihood based boosting), "skip" (no filter), "kendall" (stats::cor.test), "spearman" (stats::cor.test), "pearson" (stats::cor.test)
<code>until</code>	Stop at index/column (if 0: iterate through all columns). For testing purposes in large datasets.
<code>progress</code>	Integer. If > 0, print progress after every X steps (Progress might not be reported completely accurate due to parallel execution)
<code>mode</code>	Integer. Mode (0: x86, 1: FMA, 2: AVX). Features are only available if compiled accordingly and available on the hardware.
<code>soft_power</code>	Integer. Exponent of the transformation. Set automatically based on the scale free topology criterion if unspecified.
<code>max_singleton</code>	Integer. The maximal singleton in the clustering. Usually equals the number of features.
<code>qc_plot</code>	Logical. Should plots be created?
<code>min_cluster_size</code>	Integer. The minimum number of features in one module.
<code>ME_diss_thres</code>	Numeric. Module Eigengene Dissimilarity Threshold for merging close modules.

n_pc	Number of principal components and variance explained entries to be calculated. The number of returned variance explained entries is currently 'min(n_pc,10)'. If given 'n_pc' is greater than 10, a warning is issued.
robust_PCs	Should PCA be calculated on ranked data (Spearman PCA)? Rotations will not correspond to original data if this is applied.
nb_min_varExpl	Minimum proportion of variance explained for returned module eigengenes. The number of PCs is capped at n_pc.
cores	Integer. Amount of CPU cores used (<=1 : sequential)
scale	Logical. Should data be scaled and centered?
method	A character string specifying the method to be used for correlation coefficients.
verbose	Additional diagnostic messages.

Value

dendros A list of dendrograms. For each fully separate part of the network an individual dendrogram.

names A vector of feature names.

colors A vector of numeric color coding in matching order of names and module eigengene names (color = 3 -> variable in ME3).

MEs Aggregated module measures (Module eigengenes).

var_explained Proportion of variance explained per module eigengene per principal component (max n_pc principal components are listed).

rotation Matrix of variable loadings divided by their singular values. datan

filter Filter-Matrix as generated by the nb_filter function.

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
results <- netboost(datan=tcga_aml_meth_rna_chr18, stepno=20L,
  soft_power=3L, min_cluster_size=10L, n_pc=2, scale=TRUE,
  ME_diss_thres=0.25, qc_plot=TRUE)
```

netboostMCUPGMAPath	<i>Returns the absolute path to folder with mcupgma executables and scripts.</i>
---------------------	--

Description

Returns the absolute path to folder with mcupgma executables and scripts.

Usage

```
netboostMCUPGMAPath()
```

Value

Absolute path for "mcupgma" folder

netboostPackagePath	Returns the absolute path to "exec" folder in the package.
---------------------	--

Description

Returns the absolute path to "exec" folder in the package.

Usage

```
netboostPackagePath()
```

Value

Absolute path of installed package

netboostTmpCleanup	Cleans the netboost temporary folder. This can be useful during the session as mcupgma creates vast directory structures (for iterations). Creates the own folder (all netboost temporary data is stored in netboostTmpPath(), which is equal to tempdir()/netboost). Also used for first time setup of folder.
--------------------	---

Description

Cleans the netboost temporary folder. This can be useful during the session as mcupgma creates vast directory structures (for iterations). Creates the own folder (all netboost temporary data is stored in netboostTmpPath(), which is equal to tempdir()/netboost). Also used for first time setup of folder.

Usage

```
netboostTmpCleanup(verbose = FALSE)
```

Arguments

verbose	Flag verbose
---------	--------------

Value

none

netboostTmpPath	<i>Returns the absolute path to temporary folder of the package. To change temporary path, use normal R variables (TEMPDIR etc).</i>
-----------------	--

Description

Returns the absolute path to temporary folder of the package. To change temporary path, use normal R variables (TEMPDIR etc).

Usage

```
netboostTmpPath()
```

Value

Absolute path for temporary folder

tcga_aml_meth_rna_chr18	<i>TCGA RNA and methylation measurement on a subset of chromosome 18 for 80 AML patients.</i>
-------------------------	---

Description

TCGA RNA and methylation measurement on a subset of chromosome 18 for 80 AML patients.

Usage

```
tcga_aml_meth_rna_chr18
```

Format

A data frame with 80 rows (patients) and 500 variables (features).

Source

<https://portal.gdc.cancer.gov/>

tree_dendro	<i>Calculate the dendrogram for an individual tree</i>
-------------	--

Description

Calculate the dendrogram for an individual tree

Usage

```
tree_dendro(tree, datan, forest)
```

Arguments

tree	A list with two elements. ids, which is an integer vector of feature identifiers and rows, which is an integer vector of selected rows in the corresponding forest
datan	Data frame where rows correspond to samples and columns to features.
forest	Raw dendrogram-matrix as generated by the nb_mcupgma function.

Value

List of tree specific objects including dendrogram, tree data and features.

tree_search	<i>Extracts independent trees from nb_mcupgma results (external wrapper for internal C++ function)</i>
-------------	--

Description

Extracts independent trees from nb_mcupgma results (external wrapper for internal C++ function)

Usage

```
tree_search(forest = NULL)
```

Arguments

forest	Raw dendrogram-matrix as generated by the nb_mcupgma function.
--------	--

Value

List

Examples

```
data('tcga_aml_meth_rna_chr18', package='netboost')
cores <- as.integer(getOption('mc.cores', 2))
datan <- as.data.frame(scale(tcga_aml_meth_rna_chr18, center=TRUE,
                             scale=TRUE))
filter <- nb_filter(datan=datan, stepno=20L, until=0L, progress=1000L,
                    cores=cores, mode=2L)
dist <- nb_dist(datan=datan, filter=filter, soft_power=3L, cores=cores)
max_singleton = dim(tcga_aml_meth_rna_chr18)[2]
forest <- nb_mcupgma(filter=filter, dist=dist,
                     max_singleton=max_singleton, cores=cores)
trees <- tree_search(forest)
str(trees[[length(trees)]])
```

Index

[.onAttach](#), [3](#)

[calculate_adjacency](#), [3](#)

[cpp_dist_tom](#), [4](#)

[cpp_filter_base](#), [5](#), [6](#)

[cpp_filter_end](#), [5](#)

[cpp_filter_step](#), [6](#)

[cpp_tree_search](#), [6](#)

[cut_dendro](#), [7](#)

[cut_trees](#), [8](#)

[cutreeDynamic](#), [12](#)

[mcupgma_exec](#), [9](#)

[nb_clust](#), [10](#)

[nb_consensus](#), [11](#), [13](#), [14](#)

[nb_consensus_transfer](#), [13](#)

[nb_dist](#), [14](#)

[nb_filter](#), [12](#), [15](#)

[nb_mcupgma](#), [16](#)

[nb_moduleEigengenes](#), [17](#)

[nb_plot_dendro](#), [20](#)

[nb_summary](#), [21](#)

[nb_transfer](#), [21](#)

[netboost](#), [22](#)

[netboostMCUPGMAPath](#), [24](#)

[netboostPackagePath](#), [25](#)

[netboostTmpCleanup](#), [25](#)

[netboostTmpPath](#), [26](#)

[tcga_aml_meth_rna_chr18](#), [26](#)

[tree_dendro](#), [27](#)

[tree_search](#), [27](#)