Package 'RAFS'

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Description A cross-validated minimal-optimal feature selection algorithm. It utilises popularity counting, hierarchical clustering with feature dissimilarity measures, and prefiltering with all-relevant feature selection method to obtain the minimal-optimal set of features.				
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 $builtin_dist_funs$

All built-in feature dissimilarity functions

Description

To be used in run_rafs.

Usage

builtin_dist_funs

Format

An object of class list of length 5.

Details

See also default_dist_funs.

compute_fs_results

Compute preliminary feature selection results for RAFS

Description

This is a secondary function, useful when experimenting with different feature selection filters and rankings. Its output is used in run_rafs_with_fs_results and it is called for the user in run_rafs.

```
compute_fs_results(data, decision, k, seeds, fs_fun = default_fs_fun)
```

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Arguments

data	input data where columns are variables and rows are observations (all numeric)
decision	decision variable as a binary sequence of length equal to number of observations
k	number of folds for internal cross validation
seeds	a vector of seeds used for fold generation for internal cross validation
fs_fun	function to compute feature selection p-values, it must have the same signature as default_fs_fun (which is the default, see its help to learn more)

Value

A list with feature selection results, e.g. from default_fs_fun.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
fs_results <- compute_fs_results(madelon$data, madelon$decision, 2, c(12345))
run_rafs_with_fs_results(madelon$data, madelon$decision, fs_results)</pre>
```

cor_dist

Feature dissimilarity based on Pearson's Correlation (cor)

Description

To be used as one of the dist_funs in run_rafs.

Usage

```
cor_dist(relevant_train_data, train_decision = NULL, seed = NULL)
```

Arguments

relevant_train_data

input data where columns are variables and rows are observations (all numeric);

assumed to contain only relevant data

train_decision decision variable as a binary sequence of length equal to number of observations

seed a numerical seed

Value

A matrix of distances (dissimilarities).

default_dist_funs

create_seeded_folds
Create seeded folds

Description

A utility function used in RAFS but useful also for external cross-validation.

Usage

```
create_seeded_folds(decision, k, seed)
```

Arguments

decision decision variable as a binary sequence of length equal to number of observations

k number of folds for cross validation

seed a numerical seed

Value

A vector of folds. Each fold being a vector of selected indices.

Description

As used in run_rafs.

Usage

```
default_dist_funs
```

Format

An object of class list of length 3.

Details

The default functions compute: Pearson's correlation (cor: cor_dist), Variation of Information (vi: vi_dist) and Symmetric Target Information Gain (stig: stig_dist).

These functions follow a similar protocol to default_fs_fun. They expect the same input except for the assumption that the data passed in is relevant. Each of them outputs a matrix of distances (dissimilarities) between features.

See also builtin_dist_funs.

default_fs_fun 5

default_fs_fun

Default (example) feature selection function for RAFS

Description

See run_rafs for how it is used. Only the train portion of the dataset is to be fed into this function.

Usage

```
default_fs_fun(train_data, train_decision, seed)
```

Arguments

train_data input data where columns are variables and rows are observations (all numeric) train_decision decision variable as a binary sequence of length equal to number of observations seed a numerical seed

Details

The function MUST use this train_data and MAY ignore the train_decision.

If the function depends on randomness, it MUST use the seed parameter to seed the PRNG.

The function needs to return a list with at least two elements: rel_vars and rel_vars_rank, which are vectors and contain, respectively, the indices of variables considered relevant and the rank for each relevant variable. The function MAY return a list with more elements.

Other examples of sensible functions are included in the tests of this package.

Value

A list with at least two fields: rel_vars and rel_vars_rank, which are vectors and contain, respectively, the indices of variables considered relevant and the rank for each relevant variable.

```
default_hclust_methods
```

Default hclust methods

Description

As used in run_rafs to call hclust.

Usage

```
default_hclust_methods
```

Format

An object of class character of length 4.

Description

This helper function works on results of get_rafs_reps_popents to obtain all representatives at the chosen number of clusters.

Usage

```
get_rafs_all_reps_from_popcnts(reps_popcnts, n_clusters)
```

Arguments

```
reps_popcnts representatives' popcnts for the chosen variant as obtained from get_rafs_reps_popcnts
n_clusters the desired number of clusters
```

Value

A vector of all representatives.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
rafs_reps_popents <- get_rafs_reps_popents(rafs_results, 5)
get_rafs_all_reps_from_popents(rafs_reps_popents$stig_single, 5)</pre>
```

```
get_rafs_occurrence_matrix
```

Get co-occurrence matrix from RAFS results

Description

This function obtains a matrix describing a graph of co-occurrence at each count of clusters (from n_clusters_range) computed over all runs of RAFS.

```
get_rafs_occurrence_matrix(
  rafs_results,
  interesting_reps,
  n_clusters_range = 2:15
)
```

get_rafs_reps_popents 7

Arguments

```
rafs_results RAFS results as obtained from run_rafs
interesting_reps
the interesting representatives to build matrices for (in principle, these need not be representatives but it is more common)
n_clusters_range
range of clusters number to obtain matrices for
```

Details

If a single result over a cluster number range is desired, the selected matrices can be summed.

Value

A nested list with matrices. The first level is per the RAFS variant (combination of feature dissimilarity function and helust method). The second level is per the number of clusters. The third (and last) level is the co-occurrence matrix.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
rafs_reps_popents <- get_rafs_reps_popents(rafs_results, 5)
rafs_top_reps <- get_rafs_top_reps_from_popents(rafs_reps_popents$stig_single, 5)
get_rafs_occurrence_matrix(rafs_results, rafs_top_reps, 5)</pre>
```

Description

This function obtains popularity counts (popcnts) of representatives present at each count of clusters (from n_clusters_range) computed over all runs of RAFS.

Usage

```
get_rafs_reps_popcnts(rafs_results, n_clusters_range = 2:15)
```

Arguments

```
rafs_results RAFS results as obtained from run_rafs
n_clusters_range
range of clusters number to obtain popents for
```

Details

These results might be fed into further helper functions: get_rafs_top_reps_from_popents and get_rafs_all_reps_from_popents.

Value

A nested list with popcnts. The first level is per the RAFS variant (combination of feature dissimilarity function and helust method). The second level is per the number of clusters. The third (and last) level is popcnts per representative.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
get_rafs_reps_popcnts(rafs_results, 2:5)</pre>
```

Description

This function obtains a matrix of representatives's describing a graph of co-representation at each count of clusters (from n_clusters_range) computed over all runs of RAFS.

Usage

```
get_rafs_rep_tuples_matrix(
  rafs_results,
  interesting_reps,
  n_clusters_range = 2:15
)
```

Arguments

```
rafs_results RAFS results as obtained from run_rafs
interesting_reps
the interesting representatives to build matrices for
n_clusters_range
range of clusters number to obtain matrices for
```

Details

If a single result over a cluster number range is desired, the selected matrices can be summed.

Value

A nested list with matrices. The first level is per the RAFS variant (combination of feature dissimilarity function and helust method). The second level is per the number of clusters. The third (and last) level is the co-representation matrix.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
rafs_reps_popents <- get_rafs_reps_popents(rafs_results, 5)
rafs_top_reps <- get_rafs_top_reps_from_popents(rafs_reps_popents$stig_single, 5)
get_rafs_rep_tuples_matrix(rafs_results, rafs_top_reps, 5)</pre>
```

```
get_rafs_rep_tuples_popcnts
```

Get representatives' tuples' popularity counts (popcnts) from RAFS results

Description

This function obtains popularity counts (popcnts) of representatives' tuples present at each count of clusters (from n_clusters_range) computed over all runs of RAFS.

Usage

```
get_rafs_rep_tuples_popcnts(rafs_results, n_clusters_range = 2:15)
```

Arguments

```
rafs_results RAFS results as obtained from run_rafs n_clusters_range
```

range of clusters number to obtain popents for

Value

A nested list with popcnts. The first level is per the RAFS variant (combination of feature dissimilarity function and helust method). The second level is per the number of clusters. The third (and last) level is popcnts per representatives' tuple.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
get_rafs_rep_tuples_popcnts(rafs_results, 2:5)</pre>
```

get_rafs_tops_popents Get top popularity counts (popents) from FS results

Description

This function obtains popularity counts (popents) of top variables computed over all runs of FS.

Usage

```
get_rafs_tops_popcnts(fs_results, n_top_range = 2:15)
```

Arguments

fs_results RAFS FS results as obtained from compute_fs_results n_top_range range of top number to obtain popents for

Details

These results might be fed into further helper functions: get_rafs_top_reps_from_popents and get_rafs_all_reps_from_popents.

Value

A nested list with popents. The first level is per the number of top variables. The second (and last) level is popents per top variable.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
fs_results <- compute_fs_results(madelon$data, madelon$decision, 2, c(12345))
get_rafs_tops_popcnts(fs_results, 2:5)</pre>
```

```
get_rafs_top_reps_from_popcnts
```

Get top (i.e., most common) representatives from their popents

Description

This helper function works on results of get_rafs_reps_popents to obtain the desired number of top (most common) representatives at the chosen number of clusters.

```
get_rafs_top_reps_from_popcnts(reps_popcnts, n_clusters, n_reps = n_clusters)
```

Arguments

reps_popcnts popcnts for the chosen variant as obtained from get_rafs_reps_popcnts
n_clusters the desired number of clusters

n_reps the desired number of top representatives

Value

A vector of top representatives.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
rafs_reps_popcnts <- get_rafs_reps_popcnts(rafs_results, 5)
get_rafs_top_reps_from_popcnts(rafs_reps_popcnts$stig_single, 5)</pre>
```

```
get_rafs_top_rep_tuples_from_popcnts
```

Get top (i.e., most common) representatives's tuples from their popents

Description

This helper function works on results of get_rafs_rep_tuples_popcnts to obtain the desired number of top (most common) representatives' tuples at the chosen number of clusters.

Usage

```
get_rafs_top_rep_tuples_from_popcnts(
  rep_tuples_popcnts,
  n_clusters,
  n_tuples = 1
)
```

Arguments

```
rep_tuples_popcnts
```

tuples' popents for the chosen variant as obtained from get_rafs_rep_tuples_popents

n_clusters the desired number of clustersn_tuples the desired number of top tuples

Value

A list of top tuples (each tuple being a vector of representatives).

run_rafs

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
rafs_results <- run_rafs(madelon$data, madelon$decision, 2, c(12345))
rafs_rep_tuples_popents <- get_rafs_rep_tuples_popents(rafs_results, 5)
get_rafs_top_rep_tuples_from_popents(rafs_rep_tuples_popents$tig_single, 5)</pre>
```

get_run_id

Generate CV run identifiers

Description

A utility function used in RAFS to generate cross validation run identifiers, thus useful also for external cross-validation.

Usage

```
get_run_id(seed, k, i)
```

Arguments

seed a numerical seed
k number of folds for cross validation
i current fold number (1 to k)

Value

A string with the run identifier.

run_rafs

Robust Aggregative Feature Selection (RAFS)

Description

This is the main function of the RAFS library to run for analysis.

```
run_rafs(
  data,
  decision,
  k = 5,
  seeds = sample.int(32767, 10),
  fs_fun = default_fs_fun,
  dist_funs = default_dist_funs,
  hclust_methods = default_hclust_methods
)
```

Arguments

data	input data where columns are variables and rows are observations (all numeric)
decision	decision variable as a binary sequence of length equal to number of observations
k	number of folds for internal cross validation
seeds	a vector of seeds used for fold generation for internal cross validation
fs_fun	function to compute feature selection p-values, it must have the same signature as default_fs_fun (which is the default, see its help to learn more)
dist_funs	a list of feature dissimilarity functions computed over the relevant portion of the training dataset (see the example $default_dist_funs$ and $builtin_dist_funs$ to learn more)
hclust_methods	a vector of hclust methods to use

Details

Depending on your pipeline, you may want to also check out run_rafs_with_fs_results and compute_fs_results which this function simply wraps over.

The results from this function can be fed into one of the helper functions to analyse them further: get_rafs_reps_popents, get_rafs_rep_tuples_popents, get_rafs_rep_tuples_matrix and get_rafs_occurrence_matrix.

Value

A nested list with hclust results. The first level is per the cross validation run. The second level is per the feature dissimilarity function. The third (and last) level is per the hclust method.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
run_rafs(madelon$data, madelon$decision, 2, c(12345))
```

```
run_rafs_with_fs_results
```

Robust Aggregative Feature Selection (RAFS) from feature selection results

Description

This is a secondary function, useful when experimenting with different feature selection filters and rankings. The output is exactly the same as from run_rafs.

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Usage

```
run_rafs_with_fs_results(
  data,
  decision,
  fs_results,
  dist_funs = default_dist_funs,
  hclust_methods = default_hclust_methods
)
```

Arguments

input data where columns are variables and rows are observations (all numeric)
decision decision variable as a binary sequence of length equal to number of observations
output from compute_fs_results computed for the same data and decision
dist_funs a list of feature dissimilarity functions computed over the relevant portion of the training dataset (see the example default_dist_funs to learn more)
hclust_methods a vector of hclust methods to use

Value

A nested list with hclust results. The first level is per the cross validation run. The second level is per the feature dissimilarity function. The third (and last) level is per the hclust method.

Examples

```
library(MDFS)
mdfs_omp_set_num_threads(1) # only to pass CRAN checks
data(madelon)
fs_results <- compute_fs_results(madelon$data, madelon$decision, 2, c(12345))
run_rafs_with_fs_results(madelon$data, madelon$decision, fs_results)</pre>
```

stig_dist

Symmetric Target Information Gain (STIG) computed directly

Description

To be used as one of the dist_funs in run_rafs.

```
stig_dist(relevant_train_data, train_decision, seed)
```

stig_from_ig_dist 15

Arguments

relevant_train_data

input data where columns are variables and rows are observations (all numeric);

assumed to contain only relevant data

train_decision decision variable as a binary sequence of length equal to number of observations

seed a numerical seed

Details

This function computes the STIG metric directly from the data, maximising it over 30 discretisations.

Value

A matrix of distances (dissimilarities).

stig_from_ig_dist

Symmetric Target Information Gain (STIG) computed from single Information Gains (IGs)

Description

To be used as one of the dist_funs in run_rafs.

Usage

```
stig_from_ig_dist(relevant_train_data, train_decision, seed)
```

Arguments

relevant_train_data

input data where columns are variables and rows are observations (all numeric);

assumed to contain only relevant data

train_decision decision variable as a binary sequence of length equal to number of observations seed a numerical seed

Details

This function computes the STIG metric from single Information Gains (IGs) maximised over 30 discretisations and then summed pair-wise.

This function is similar to stig_dist but the results differ slightly. We recommend the direct computation in general.

Value

A matrix of distances (dissimilarities).

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stig_stable_dist	Symmetric Target Information Gain (STIG) computed directly but with pre-computed 1D conditional entropy (aka stable)
	pre computed 12 conditional entropy (and state)

Description

To be used as one of the dist_funs in run_rafs.

Usage

```
stig_stable_dist(relevant_train_data, train_decision, seed)
```

Arguments

relevant_train_data

input data where columns are variables and rows are observations (all numeric); assumed to contain only relevant data

train_decision decision variable as a binary sequence of length equal to number of observations seed a numerical seed

Details

This function computes the STIG metric directly from the data, maximising it over 30 discretisations, but reusing the common 1D conditional entropy.

Value

A matrix of distances (dissimilarities).

vi_dist

Variation of Information (VI)

Description

To be used as one of the dist_funs in run_rafs.

Usage

```
vi_dist(relevant_train_data, train_decision = NULL, seed)
```

Arguments

relevant_train_data

input data where columns are variables and rows are observations (all numeric); assumed to contain only relevant data

train_decision decision variable as a binary sequence of length equal to number of observations seed a numerical seed

Details

This function computes the Variation of Information (VI) averaged over 30 discretisations.

Value

A matrix of distances (dissimilarities).

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