

Package ‘TDMR’

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Type Package

Title Tuned Data Mining in R

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Description Tuned Data Mining in R (TDMR) performs a complete tuning of a data mining task.

License GPL (>= 2)

Depends R (>= 2.14.0), SPOT, twiddler

Suggests cmaes, parallel, e1071, powell, ROCR, randomForest, rCMA,
rSFA

Imports testit, adabag

Collate 'makeTdmRandomSeed.r' 'printTDMclassifier.r'
'printTDMregressor.r' 'tdmBigLoop.r' 'tdmClassify.r'
'tdmClassifyLoop.r' 'tdmDefaultsFill.r' 'tdmDispatchTuner.r'
'tdmEmbedDataFrame.r' 'tdmEnvTMakeNew.r' 'tdmGeneralUtils.r'
'tdmGetObj.r' 'tdmGraphicUtils.r' 'tdmMapDesign.r'
'tdmMetacostRf.r' 'tdmModelingUtils.r' 'tdmOptsDefaults.r'
'tdmParaBootstrap.r' 'tdmPlotResMeta.r' 'tdmPreprocUtils.r'
'tdmReadData.r' 'tdmROCR.r' 'tdmRegress.r' 'tdmRegressLoop.r'
'tdmSplitTestData.r' 'tdmStartSpot.r' 'tdmStartOther.r'
'unbiasedBestRun_O.r' 'unbiasedRun.r' 'tdmExecSpotStep.r'

NeedsCompilation no

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TDMR-package	<i>Tuned Data Mining in R</i>
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Description

Tuned Data Mining in R

Details

Package:	TDMR
Type:	Package
Version:	1.3
Date:	30.08.2014
License:	GPL (>= 2)
LazyLoad:	yes

TDMR is a package for tuned data mining (predictive analytics, i.e. **classification** and **regression**). Its main features are:

- 1) A variety of tuners, with special emphasis on [SPOT](#) (a well-known R package for parameter tuning), but also CMA-ES (package [rCMA](#)) and other tuning algorithms.
- 2) Tuning of preprocessing (feature generation) parameters and model building parameters simultaneously.
- 3) Support for multiple tuning experiments (different settings, repetitions with different resamplings, ...).
- 4) Easy parallelization of those experiments with the help of R packages [snow](#) and [parallel](#).
- 5) Extensibility: New tuning parameters, new feature preprocessing tools, model builders and even new tuners can be added easily.

The main entry point functions are [tdmClassifyLoop](#), [tdmRegressLoop](#) and [tdmBigLoop](#). See [tdmOptsDefaultsSet](#) and [tdmDefaultsFill](#) for an overview of adjustable TDMR-parameters.

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), Patrick Koch

References

<http://gociop.de/research-projects/tuned-data-mining/>

dsetTest.TDMdata *Return the test part of a [TDMdata](#) object containing the task data.*

Description

Return the test part of a [TDMdata](#) object containing the task data.

Usage

```
## S3 method for class 'TDMdata'  
dsetTest(x)
```

Arguments

x return value from a prior call to [tdmSplitTestData](#), an object of class [TDMdata](#).

Value

tset, a data frame with all test records. If there are 0 test records, return NULL.

Author(s)

Wolfgang Konen, FHK, Feb'2012

See Also

[unbiasedRun](#) [dsetTrnVa.TDMdata](#)

dsetTrnVa.TDMdata	<i>Return the train-validation part of a TDMdata object containing the task data.</i>
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Description

Return the train-validation part of a [TDMdata](#) object containing the task data.

Usage

```
## S3 method for class 'TDMdata'  
dsetTrnVa(x)
```

Arguments

x return value from a prior call to [tdmSplitTestData](#), an object of class [TDMdata](#).

Value

dset, a data frame with all train-validation records

Author(s)

Wolfgang Konen, FHK, Feb'2012

See Also

[dsetTest.TDMdata](#) [tdmSplitTestData](#)

Opts	<i>Return the list 'opts'.</i>
------	--------------------------------

Description

Returns the list opts from objects of class [TDMenvir](#), [TDMclassifier](#), [TDMregressor](#), [tdmClass](#) or [tdmRegre](#).

Usage

```
Opts(x, ...)  
  
## S3 method for class 'TDMenvir'  
Opts(x, ...)  
  
## S3 method for class 'TDMclassifier'  
Opts(x, ...)
```

```
## S3 method for class 'TDMregressor'
Opts(x, ...)

## S3 method for class 'tdmClass'
Opts(x, ...)

## S3 method for class 'tdmRegre'
Opts(x, ...)

## Default S3 method:
Opts(x, ...)
```

Arguments

x an object of class [TDMclassifier](#), [tdmClass](#), [TDMregressor](#) or [tdmRegre](#).
 ... – currently not used –

Value

the list opts with DM-specific settings contained in the specified object

predict.TDMenvir *Make a prediction using the last model.*

Description

Make a prediction with objects of class [TDMenvir](#), [TDMclassifier](#), [TDMregressor](#). The prediction is based on the (last) model trained during [unbiasedRun](#).

Usage

```
## S3 method for class 'TDMenvir'
predict(object, ...)

## S3 method for class 'TDMclassifier'
predict(object, ...)

## S3 method for class 'TDMregressor'
predict(object, ...)
```

Arguments

object an object of class [TDMenvir](#), [TDMclassifier](#), [TDMregressor](#) containing in element lastModel the relevant model.
 ... arguments passed on to the model's predict function. Usually the first argument of ... should be newdata, a data frame for which new predictions are desired.

Value

a vector with length `nrow(newdata)` containing the new predictions.

Examples

```
## Not run:
## This example requires that demo04cpu.r is executed first (it will write demo04cpu.RData)
path <- paste(find.package("TDMR"), "demo01cpu/", sep="/");
tdm <- list( filenameEnvT="demo04cpu.RData" ); # file with environment envT
load(paste(path,tdm$filenameEnvT, sep="/"));

# take only the first 15 records:
newdata=read.csv2(file=paste(path,"data/cpu.csv", sep=""), dec=".")[1:15,];
z=predict(envT,newdata);
print(z);

## End(Not run)
```

`print.TDMclassifier` *Print an overview for a [TDMclassifier](#) or [tdmClass](#) object.*

Description

Print an overview for a [TDMclassifier](#) or [tdmClass](#) object.

Usage

```
## S3 method for class 'TDMclassifier'
print(x, ...)

## S3 method for class 'tdmClass'
print(x, ...)
```

Arguments

<code>x</code>	an object of class tdmClass , as returned from a prior call to tdmClassify , or an object of class TDMclassifier , as returned from a prior call to tdmClassifyLoop .
<code>...</code>	e.g. 'type' which information to print: "overview" (default) relative gain on training/test set, number of records, see tdmClassifySummary "cm.train" confusion matrix on train set "cm.vali" confusion matrix on test set "?" help on this method

Author(s)

Wolfgang Konen, FHK, 2011 - 2013

See Also

[tdmClassify](#), [tdmClassifySummary](#), [TDMclassifier](#)

print.TDMdata *Print an overview for a [TDMdata](#) object.*

Description

Print number of rows and number of columns of the data frame dset contained in the [TDMdata](#) object.

Usage

```
## S3 method for class 'TDMdata'
print(x, ...)
```

Arguments

x return value from a prior call to [tdmSplitTestData](#), an object of class [TDMdata](#).
 ... currently not used

Author(s)

Wolfgang Konen, FHK

See Also

[tdmSplitTestData](#)

print.TDMregressor *Print an overview for a [TDMregressor](#) or [tdmRegre](#) object.*

Description

Print an overview for a [TDMregressor](#) or [tdmRegre](#) object.

Usage

```
## S3 method for class 'TDMregressor'
print(x, ...)
```

```
## S3 method for class 'tdmRegre'
print(x, ...)
```


Arguments

- x an object of class [tdmRegre](#), as returned from a prior call to [tdmRegress](#),
or an object of class [TDMregressor](#), as returned from a prior call to [tdmRegressLoop](#).
- ... e.g. 'type' which information to print:
"overview" (def.) RMAE on training/test set, number of records, see [tdmRegressSummary](#)
"... " ... other choices, TODO ...
"?" help on this method

See Also

[tdmRegress](#), [tdmRegressSummary](#), [TDMregressor](#)

Tuning and unbiased evaluation in a big loop.

Description

For each .conf file in `tdm$runList` call all tuning algorithms (SPOT, CMA-ES or other) specified in `tdm$tuneMethod` (via function [tdmDispatchTuner](#)). After each tuning process perform a run of `tdm$unbiasedFunc` (usually [unbiasedRun](#)).

Each of these experiments is repeated `tdm$nExperim` times. Thus we have for each triple

$$(confFile, nExp, theTuner)$$

a tuning result. The ranges of the triple elements are:

```
confFile in tdm$runList
nExp in 1,...,tdm$nExperim
theTuner in tdm$tuneMethod
```

Usage

```
tdmBigLoop(envT, spotStep = "auto", dataObj = NULL)
```

Arguments

envT an environment containing on input at least the element `tdm` (a list with general settings for TDMR, see [tdmDefaultsFill](#)), which has at least the elements
`tdm$runList` vector of .conf filenames
`tdm$spotList` [NULL] vector of .conf filenames for which spot tuning is done.
If NULL, then `spotList=runList`

spotStep	["auto"] which step of SPOT to execute (either "auto" or "rep").
dataObj	[NULL] optional object of class <code>TDMdata</code> (the same for all runs in big loop). If it is NULL, it will be constructed here with the help of <code>tdmSplitTestData</code> . Then it may be different for each run in the big loop.

Details

tdm refers to `envT$tdm`.

Tuning is skipped if the `.conf` file does not appear in `tdm$spotList` or if `spotStep=="rep"`. In this case it is assumed then that `envT$bstGrid` and `envT$resGrid` contain the appropriate data frames already. See `tdmEnvTAddBstRes` on how to fill `envT$bstGrid` and `envT$resGrid` from an `.RData` file.

The available tuning algorithms (tuners) are

- `spotTuner`: Call `spot`.
- `lhdTuner`: Perform a parameter tuning using a Latin hypercube design (LHD) for obtaining best design points. LHD is performed by configuring SPOT in such a way that all the budget is used for the initial design (usually LHD).
- `cma_jTuner`: Perform a parameter tuning by CMA-ES, using the **Java** implementation by Niko Hansen through the interface package `rCMA`.
- `cmaesTuner`: Perform a parameter tuning by CMA-ES, using the **R**-implementation (package `cmaes` by Olaf Mersmann) (deprecated, use `cma_jTuner` instead).
- `bfgsTuner`: Perform a parameter tuning by Broyden, Fletcher, Goldfarb and Shanno (BFGS) method. The L-BFGS-B version allowing box constraints is used.
- `powellTuner`: Perform a parameter tuning by Powell's UObyQA algorithm (unconstrained optimization by quadratic approximation), see package `powell`).

Value

environment `envT`, containing the results

<code>res</code>	data frame with results from last tuning (one line for each call of <code>tdmStart*</code>)
<code>bst</code>	data frame with the best-so-far results from last tuning (one line collected after each (SPO) step)
<code>resGrid</code>	list with data frames <code>res</code> from all tuning runs. Use <code>envT\$getRes(envT, confFile, nExp, theTuner)</code> to retrieve a specific <code>res</code> .
<code>bstGrid</code>	list with data frames <code>bst</code> from all tuning runs. Use <code>envT\$getBst(envT, confFile, nExp, theTuner)</code> to retrieve a specific <code>bst</code> .
<code>theFinals</code>	data frame with one line for each triple (<code>confFile, nExp, tuner</code>), each line contains summary information about the tuning run in the form: <code>confFile tuner nExp [params] NRUN NEVAL RGain.bst RGain.* sdR.*</code> where <code>[params]</code> is written depending on <code>tdm\$withParams</code> . <code>NRUN</code> is the number of unbiased evaluation runs.

NEVAL is the number of function evaluations (model builds) during tuning. RGain denotes the relative gain on a certain data set: the actual gain achieved with the model divided by the maximum gain possible for the current cost matrix and the current data set. This is for classification tasks, in the case of regression each `RGain.*` is replaced by `RMAE.*`, the relative mean absolute error. Each `'sdR.'` denotes the standard deviation of the preceding RGain or RMAE. `RGain.best` is the best result during tuning obtained on the training-validation data. `RGain.avg` is the average result during tuning. The following pairs `RGain.*sdR.*` are the results of one or several unbiased evaluations on the test data where `'*'` takes as many values as there are elements in `tdm$umode` (the possible values are explained in [unbiasedRun](#)).

result	object of class <code>TDMclassifier</code> or <code>TDMregressor</code> . This is a list with results from <code>tdm\$mainFunc</code> as called in the last unbiased evaluation using the best parameters found during tuning. Use <code>print(envT\$result)</code> to get more info on such an object of class <code>TDMclassifier</code> .
tunerVal	an object with the return value from the last tuning process. For every tuner, this is the list <code>spotConfig</code> , containing the SPOT settings plus the TDMR settings in elements <code>opts</code> and <code>tdm</code> . Every tuner extends this list by <code>tunerVal\$alg.currentResult</code> and <code>tunerVal\$alg.currentBest</code> , see tdmDispatchTuner . In addition, each tuning method might add specific elements to the list, see the description of each tuner.

Environment `envT` contains further elements, but they are only relevant for the internal operation of `tdmBigLoop` and its subfunctions.

Note

Side effects: Irrespective of the value of `tdm$fileMode`,

- a compressed version of `envT` is saved to file `tdm$filenameEnvT` (default: `<runList[1]>.RData`), relative to the directory of the `.conf` file.

If `tdm$U.saveModel==TRUE`, then `envT$result$lastRes$lastModel` (the last trained model) will be saved to `tdm$filenameEnvT`. The default is `tdm$U.saveModel==FALSE` (smaller `.RData` files).

If `tdm$fileMode==TRUE`, more files are written relative to the directory of the `.conf` file:

- `envT$theFinals` is written to file `tdm$finalFile`
- `envT$theFinals` is appended to `tdm$experFile`

If `tdm$finalFile==NULL`, then it is set to `sub(".conf", ".fin", runList[1])`.

If `tdm$experFile==NULL`, then nothing is appended to any experiment file.

Example usages of function `tdmBigLoop` are shown in

```
demo(demo03sonar)
demo(demo03sonar_B)
demo(demo04cpu)
```

where the corresponding R-sources are in directory demo.

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), Patrick Koch

See Also

[tdmDispatchTuner](#), [unbiasedRun](#)

Examples

```

### This demo shows a complete tuned data mining process (level 3 of TDMR) where
### the data mining task is the classification task SONAR (from UCI repository,
### http://archive.ics.uci.edu/ml/datasets/Connectionist+Bench+%28Sonar,+Mines+vs.+Rocks%29).
### The data mining process is in main_sonar.r, which calls tdmClassifyLoop and tdmClassify
### with Random Forest as the prediction model.
### The three parameter to be tuned are CUTOFF1, CLASSWT2 and XPERC, as specified
### in file sonar_04.roi. The tuner used here is LHD.
### Tuning runs are rather short, to make the example run quickly.
### Do not expect good numeric results.
### See demo/demo03sonar_B.r for a somewhat longer tuning run, with two tuners SPOT and LHD.

## set working directory (dir with .apd, .conf and main_*.r file)
path <- paste(find.package("TDMR"), "demo02sonar", sep="/");
source(paste(path,"main_sonar.r", sep="/"));

## control settings for TDMR
tdm <- list( mainFunc="main_sonar"
            , runList = c("sonar_04.conf")
            , umode="CV" # { "CV" | "RSUB" | "TST" | "SP_T" }
            , tuneMethod = c("lhd")
            , filenameEnvT="exBigLoop.RData" # file to save environment envT (in dir 'path')
            , nrun=1, nfold=2 # repeats and CV-folds for the unbiased runs
            , nExperim=1
            , parallelCPUs=1
            , parallelFuncs=c("readCmdSonar")
            , optsVerbosity = 0 # the verbosity for the unbiased runs
            );

## Each element of tdm$runList has the settings for one tuning process (e.g.
## - auto.loop.steps = number of SPOT generations
## - auto.loop.evals = budget of model building runs and
## - io.roiFileName = "sonar_04.roi"
## ).

spotStep = "auto";
source(paste(path,"start_bigLoop.r", sep="/"), chdir=TRUE); # change dir to 'path' while sourcing

```

tdmBindResponse	<i>Bind a column to a data frame.</i>
-----------------	---------------------------------------

Description

Bind the column with name `response.predict` and contents `vec` as last column to data frame `d`

Usage

```
tdmBindResponse(d, response.predict, vec)
```

Arguments

<code>d</code>	data frame
<code>response.predict</code>	name of new column
<code>vec</code>	the contents for the last column bound to data frame <code>d</code>

Value

data frame `d` with column added

tdmClassify	<i>Core classification function of TDMR.</i>
-------------	--

Description

`tdmClassify` is called by `tdmClassifyLoop` and returns an object of class `tdmClass`. It trains a model on training set `d_train` and evaluates it on test set `d_test`. If this function is used for tuning, the test set `d_test` plays the role of a validation set.

Usage

```
tdmClassify(d_train, d_test, d_dis, d_preproc, response.variables,
            input.variables, opts)
```

Arguments

<code>d_train</code>	training set
<code>d_test</code>	validation set, same columns as training set
<code>d_dis</code>	'disregard set', i.e. everything what is neither train nor test. The model is applied to all records in <code>d_dis</code> (needed for active learning, see <code>ssl_methods.r</code>)
<code>d_preproc</code>	data used for preprocessing. May be <code>NULL</code> , if no preprocessing is done (<code>opts\$PRE.SFA=="none"</code> and <code>opts\$PRE.PCA=="none"</code>). If preprocessing is done, then <code>d_preproc</code> is usually all non-validation data.

`response.variables` name of column which carries the target variable - or - vector of names specifying multiple target columns (these columns are not used during prediction, only for evaluation)

`input.variables` vector with names of input columns

`opts` additional parameters [defaults in brackets]

SRF.* several parameters for [tdmModSortedRFimport](#)

RF.* several parameters for RF (Random Forest, defaults are set, if omitted)

SVM.* several parameters for SVM (Support Vector Machines, defaults are set, if omitted)

filename

data.title

MOD.method ["RF"] the main training method ["RF"|"MC.RF"|"SVM"|"NB"]:
use [Random forest| MetaCost-RF| SVM| Naive Bayes] for the main model

MOD.SEED =NULL: get a new random number seed with [tdmRandomSeed](#) (different RF trainings).
=any value: set the random number seed to this value (+i) to get reproducible random numbers. In this way, the model training part (RF, NNET, ...) gets always a fixed seed (see also TST.SEED in [tdmClassifyLoop](#))

CLASSWT class weights (NULL, if all classes should have the same weight) (currently used only by methods RF, MC.RF and by [tdmModSortedRFimport](#))

fct.postproc [NULL] name of user-def'd function for postprocessing of predicted output

GD.DEVICE if !="non", then make a pairs-plot of the 5 most important variables and make a true-false bar plot

VERBOSE [2] =2: most printed output, =1: less, =0: no output

Details

Currently `d_dis` is allowed to be a 0-row data frame, but `d_train` and `d_test` must have at least one record.

Value

`res`, an object of class `tdmClass`, this is a list containing

`d_train` training set + predicted class column(s)

`d_test` test set + predicted class column(s)

`d_dis` disregard set + predicted class column(s)

`avgEVAL` list with evaluation measures, averaged over all response variables

`allEVAL` data frame with evaluation measures, one row for each response variable

`lastCmTrain` a list with evaluation info for training set (confusion matrix, gain, class errors, ...)

lastCmVali	a list with evaluation info for validation set (confusion matrix, gain, class errors, ...)
lastModel	the last model built (i.e. for the last response variable)
lastProbs	a list with three probability matrices (row: records, col: classes) v_train, v_test, v_dis, if the model provides probabilities; NULL else.
lastPred	name of the column where the prediction of the last model is appended to the datasets d_train, d_test and d_dis
predProb	a list with two data frames Trn and Val. They contain at least a column IND.dset (index of each train / validation record into data frame dset). If the model has probabilities, then they contain in addition a column for each response variable with the prediction probabilities.
opts	parameter list from input, some default values might have been added

The 9 evaluation measures in avgEVAL and alleVAL are cerr.* (misclassification error), gain.* (total gain) and rgain.* (relative gain, i.e. total gain divided by max. achievable gain in *) where * = [trn | tst | tst2] stands for [training set | test set | test set with special treatment] and the special treatment is either opts\$test2.string = "no postproc" or = "default cutoff".

The five items lastCmTrain, lastCmVali, lastModel, lastProbs, lastPred are specific for the *last* model (the one built for the last response variable in the last run and last fold)

Author(s)

Wolfgang Konen, FHK, Sep'2009 - Jun'2012

See Also

[print.tdmClass](#) [tdmClassifyLoop](#) [tdmRegressLoop](#)

Examples

```
### This demo shows a simple data mining process (phase 1 of TDMR) for classification on
### dataset iris.
### The data mining process in tdmClassify calls randomForest as the prediction model.
### It is called opts$NRUN=1 time with one random train-validation set splits.
### Therefore data frame res$allEval has one row
###
opts=tdmOptsDefaultsSet()           # set all defaults for data mining process
gdObj <- tdmGraAndLogInitialize(opts); # init graphics and log file

data(iris)
response.variables="Species"         # names, not data (!)
input.variables=setdiff(names(iris),"Species")
opts$NRUN=1

idx_train = sample(nrow(iris))[1:110]
d_train=iris[idx_train,]
d_vali=iris[-idx_train,]
d_dis=iris[numeric(0),]
res <- tdmClassify(d_train,d_vali,d_dis,NULL,response.variables,input.variables,opts)
```

```
cat("\n")
print(res$allEVAL)
```

tdmClassifyLoop	<i>Core classification double loop of TDMMR returning a TDMclassifier object.</i>
-----------------	---

Description

tdmClassifyLoop contains a double loop (opts\$NRUN and CV-folds) and calls [tdmClassify](#). It is called by all classification R-functions main_*. It returns an object of class [TDMclassifier](#).

Usage

```
tdmClassifyLoop(dset, response.variables, input.variables, opts, tset = NULL)
```

Arguments

dset	the data frame containing training and validation data.
response.variables	name of column which carries the target variable - or - vector of names specifying multiple target columns (these columns are not used during prediction, only for evaluation)
input.variables	vector with names of input columns
opts	a list from which we need here the following entries NRUN number of runs (outer loop) TST.SEED =NULL: get a new random number seed with tdmRandomSeed . =any value: set the random number seed to this value to get reproducible random numbers and thus reproducible training-test-set-selection. (only relevant in case TST.kind=="cv" or "rand") (see also MOD.SEED in tdmClassify) TST.kind how to create cvi, handed over to tdmModCreateCVindex . If TST.kind="col", then cvi is taken from dset[,opts\$TST.col]. GD.RESTART [TRUE] =TRUE/FALSE: do/don't restart graphic devices GD.DEVICE ["non" "win" "pdf" "png"]
tset	[NULL] If not NULL, this is the validation data set. If NULL, the validation data set is build from dset according to the procedure prescribed in opts\$TST.*.

Value

result, an object of class `TDMclassifier`, this is a list with results, containing

lastRes	last run, last fold: result from <code>tdmClassify</code>
C_train	classification error on training set
G_train	gain on training set
R_train	relative gain on training set (percentage of max. gain on this set)
*_vali	— similar, with vali set instead of training set —
*_vali2	— similar, with vali2 set instead of training set —
Err	a data frame with as many rows as <code>opts\$NRUN</code> and 9 columns corresponding to the nine variables described above
predictions	last run: data frame with dimensions <code>[nrow(dset),length(response.variable)]</code> . In case of CV, all CV predictions (for each record in <code>dset</code>), in other cases mixed validation / train set predictions.
predProbList	the <code>ith</code> element in this list has data on the prediction probabilities of the <code>ith</code> run. See info on <code>predProb</code> in <code>tdmClassify</code> .

Each performance measure `C_*`, `G_*`, `R_*` is a vector of length `opts$NRUN`. To be specific, `C_train[i]` is the classification error on the training set from the `i`-th run. This error is `mean(res$allEVAL$cerr.trn)`, i.e. the mean of the classification errors from all response variables when `res` is the return value of `tdmClassify`. In the case of cross validation, for each performance measure an additional averaging over all folds is done.

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), FHK, Sep'2010 - Jun'2012

See Also

`print.TDMclassifier`, `tdmClassify`, `tdmRegress`, `tdmRegressLoop`

Examples

```
### ----- demo/demo00-0classif.r -----
### This demo shows a simple data mining process (phase 1 of TDMR) for classification on
### dataset iris.
### The data mining process in tdmClassifyLoop calls randomForest as the prediction model.
### It is called opts$NRUN=2 times with different random train-validation set splits.
### Therefore data frame result$Err has two rows
###
opts=tdmOptsDefaultsSet()           # set all defaults for data mining process
gdObj <- tdmGraAndLogInitialize(opts); # init graphics and log file

data(iris)
response.variables="Species"         # names, not data (!)
input.variables=setdiff(names(iris),"Species")

result = tdmClassifyLoop(iris,response.variables,input.variables,opts)
```

```
print(result$Err)
```

tdmClassifySummary	<i>Print summary output for result from tdmClassifyLoop and add result\$y.</i>
--------------------	--

Description

result\$y is "minus OOB rgain" on training set for methods RF or MC.RF. result\$y is "minus rgain" on test set (=validation set) for all other methods. result\$y is the quantity which the tuner seeks to minimize.

Usage

```
tdmClassifySummary(result, opts, dset = NULL)
```

Arguments

result	return value from a prior call to tdmClassifyLoop , an object of class TDMclassifier.
opts	a list from which we need here the following entries NRUN number of runs (outer loop) method VERBOSE dset [NULL] if !=NULL, attach it to result
dset	[NULL] if not NULL, add this data frame to the return value (may cost a lot of memory!)

Value

result, an object of class TDMclassifier, with result\$y, result\$sd.y (and optionally also result\$dset) added

Author(s)

Wolfgang Konen, FHK, Sep'2010 - Oct'2011

See Also

[tdmClassify](#), [tdmClassifyLoop](#), [print.TDMclassifier](#), [tdmRegressSummary](#)

tdmDefaultsFill	<i>Default values for list tdm. This list controls the tuning and unbiased evaluation phase.</i>
-----------------	--

Description

When called with `tdm = tdmDefaultsFill()`, a new list `tdm` is created and returned. When called with `tdm = tdmDefaultsFill(mainFile="my.r")`, a new list `tdm` is created and returned, with the element `mainFile` set to the specified value. When called with `tdm = tdmDefaultsFill(tdm)`, an existing list `tdm` is filled with further default values.

Usage

```
tdmDefaultsFill(tdm = NULL, mainFile = NULL)
```

Arguments

<code>tdm</code>	(optional)
<code>mainFile</code>	(optional) if given, create or overwrite <code>tdm\$mainFile</code> with this value

Details

If `tdm$mainFunc` is missing, but `tdm$mainFile` exists, then `tdmDefaultsFill` will set

```
tdm$mainFunc=sub(".r","",basename(tdm$mainFile),fixed=TRUE)
```

Value

`tdm` the new / extended list, where additional elements, if they are not yet def'd, are set as:

<code>mainFile</code>	[NULL] if not NULL, source this file from the current dir. It should contain the definition of <code>tdm\$mainFunc</code> .
<code>mainFunc</code>	<code>sub(".r","",basename(tdm\$mainFile),fixed=TRUE)</code> , if <code>tdm\$mainFile</code> is set and <code>tdm\$mainFunc</code> is NULL, else "mainFunc" This is the name of the function called in tdmStartSpot and unbiasedRun
<code>unbiasedFunc</code>	["unbiasedRun"] which function to call for unbiased evaluation
<code>tuneMethod</code>	["spot"] other choices: "cmaes", "bfgs", ..., see tdmDispatchTuner
<code>nExperim</code>	[1]
<code>umode</code>	["RSUB"], one out of ["RSUB" "CV" "TST" "SP_T"], see unbiasedRun
<code>timeMode</code>	[1] 1: proc time, 2: system time, 3: elapsed time (columns <code>Time.TST</code> and <code>Time.TRN</code> in <code>envT\$theFinals</code>)
<code>fileMode</code>	[FALSE] see "Note" section in tdmBigLoop
<code>finalFile</code>	[NULL] filename where to save <code>envT\$theFinals</code> , only relevant for <code>tdm\$fileMode==TRUE</code>
<code>experFile</code>	[NULL] filename where to append <code>envT\$theFinals</code> , only relevant for <code>tdm\$fileMode==TRUE</code>

filenameEnvT	[NULL] filename where <code>tdmBigLoop</code> will save a small version of environment envT. If NULL, save envT to <code>sub(".conf", ".RData", tdm\$runList[1])</code> . This RData file is written irrespective of <code>fileMode</code> 's value, but only in case <code>spotStep=="auto"</code> .
theSpotPath	[NA] use SPOT's package version
parallelCPUs	[1] 1: sequential, >1: parallel execution with this many CPUs (package parallel)
parallelFuncs	[NULL] in case <code>tdm\$parallelCPUs>1</code> : a string vector with functions which are clusterExport'ed in addition to <code>tdm\$mainFunc</code> .
path	[NULL] where to search <code>.conf</code> and <code>.apd</code> file. If it is NULL, path is set to the actual working directory at the time when <code>tdmEnvTMakeNew</code> is executed
runList	[NULL] a list of <code>.conf</code> filenames
spotList	[NULL] a list of <code>.conf</code> filenames
stratified	[NULL] see <code>tdmSplitTestData</code>
tdmPath	[NULL] from where to source the R sources. If NULL load library TDMR instead.
test2.string	["default cutoff"]
optsVerbosity	[0] the verbosity for the unbiased runs
withParams	[TRUE] list the columns with tuned parameter in final results
nrun	[5] number of runs for unbiased run
U.saveModel	[FALSE] if TRUE, save the last model, which is trained in <code>unbiasedRun</code> , onto <code>filenameEnvT</code>
tstCol	["TST"] <code>opts\$TST.COL</code> for unbiased runs (only for <code>umode="TST"</code>)
nfold	[10] number of CV-folds for unbiased runs (only for <code>umode="CV"</code>)
TST.trnFrac	[NULL] train set fraction (of all train-vali data), OVERWRITES <code>opts\$TST.trnFrac</code> if not NULL.
TST.valiFrac	[NULL] validation set fraction (of all train-vali data), OVERWRITES to <code>opts\$TST.valiFrac</code> if not NULL.
TST.testFrac	[0.2] test set fraction (of *all* data) for unbiased runs (only for <code>umode="RSUB"</code> or <code>"SP_T"</code>)
CMA.propertyFile	[NULL] (only for CMA-ES Java tuner) see cma_jTuner .
CMA.populationSize	[NULL] (only for CMA-ES Java tuner) see cma_jTuner .

Note

The settings `tdm$TST.trnFrac` and `tdm$TST.valiFrac` allow to set programmatically certain values for `opts$TST.trnFrac` and `opts$TST.valiFrac` *after* `opts` has been read from APD file. So use `tdm$TST.trnFrac` and `tdm$TST.valiFrac` with CAUTION!

For `tdm$timeMode`, the 'user time' is the CPU time charged for the execution of user instructions of the calling process. The 'system time' is the CPU time charged for execution by the system on behalf of the calling process. The 'elapsed time' is the 'real' (wall-clock) time since the process was started.

Author(s)

Wolfgang Konen, Patrick Koch, 2011 - 2013

tdmEmbedDataFrame *Embed columns in a data frame.*

Description

Embeds the columns named `cols` in the data frame `x` into a space of dimension `dimension`.

Usage

```
tdmEmbedDataFrame(x, cols = NULL, dimension = 1, take = 1)
```

Arguments

<code>x</code>	The data frame containing the columns to embed.
<code>cols</code>	[NULL] A vector with a list of names of the columns to embed. If NULL, embed all columns of <code>x</code> .
<code>dimension</code>	[1] The additional dimensions to generate when embedding.
<code>take</code>	[1] Take from these dimensions only every <code>take</code> 'th column.

Value

The data frame, augmented with embedded columns, shortended by `dimension` rows.

tdmEnvTAddBstRes *Add BST and RES data frames to an existing envT environment.*

Description

Load an `envT`-type environment from file `fileRData`. Its elements `bst`, `bstGrid` `res`, and `resGrid` overwrite the elements in `envT` passed in as argument.

Usage

```
tdmEnvTAddBstRes(envT, fileRData)
```

Arguments

<code>envT</code>	the TDMR environment
<code>fileRData</code>	string with filename to load. This file is searched in <code>envT\$tdm\$path</code> .

Value

the augmented `envT`

tdmEnvTAddGetters	<i>Add getter functions <code>getBst</code> and <code>getRes</code> to environment <code>envT</code></i>
-------------------	--

Description

Add getter functions `getBst` and `getRes` to environment `envT`

Usage

```
tdmEnvTAddGetters(envT)
```

Arguments

<code>envT</code>	the TDMR environment
-------------------	----------------------

Value

the augmented `envT`

tdmEnvTLoad	<i>Load an <code>envT</code>-type environment from file <code>fileRData</code>.</i>
-------------	---

Description

The loaded `envT` is augmented with getter functions, see [tdmEnvTAddGetters](#).

Usage

```
tdmEnvTLoad(fileRData)
```

Arguments

<code>fileRData</code>	string with filename to load. This file is searched in the working dir.
------------------------	---

Value

`envT`

tdmEnvTMakeNew	<i>Construct a new environment envT of class TDMenvir.</i>
----------------	--

Description

Given the general TDMR settings in `tdm`, construct an appropriate environment `envT`. This is needed as input for [tdmBigLoop](#).

Usage

```
tdmEnvTMakeNew(tdm = NULL)
```

Arguments

`tdm` a list with general settings for TDMR, see [tdmDefaultsFill](#)

Value

Environment `envT`, an object of class [TDMenvir](#), containing (among others) the elements

<code>runList</code>	<code>=tdm\$runList</code>
<code>spotList</code>	<code>=tdm\$spotList</code>
<code>tdm</code>	<code>=tdmDefaultsFill(tdm)</code>
<code>getBst</code>	accessor function(<code>confFile,nExp,theTuner</code>) into <code>envT\$bstGrid</code>
<code>getRes</code>	accessor function(<code>confFile,nExp,theTuner</code>) into <code>envT\$resGrid</code>
<code>sCList</code>	list of <code>spotConfig</code> -objects, as many as <code>envT\$runList</code> has elements. Each <code>spotConfig</code> object <code>sCList[[k]]</code> contains a list <code>opts</code> as element, which is read from <code>.apd</code> file specified in <code>envT\$runList[k]</code> .

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), Patrick Koch

See Also

[tdmBigLoop](#)

tdmEnvTReadApd	<i>Re-read APD-files (list opts) for an environment envT of class TDMenvir.</i>
----------------	---

Description

Given an already existing envT with envT\$sClist filled, re-read the opts settings from APD files whose filenames are specified in envT\$sClist[[k]]\$io.apdfilename.

Usage

```
tdmEnvTReadApd(envT, tdm)
```

Arguments

envT	an environment of class TDMenvir
tdm	a list with general settings for TDMR (can be envT\$tdm)

Value

modified environment envT, an object of class [TDMenvir](#), where the items
sClist[[k]]\$opts
 =tdm\$runList
are modified

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), Patrick Koch

See Also

[tdmEnvTMakeNew](#)

tdmEnvTSensi	<i>Make a sensitivity plot based on envT</i>
--------------	--

Description

Given the results from a prior tuning run in envT, make a sensitivity plot for this run (call SPOT with spotStep="rep") If envT\$tdm\$nrnrun > 0 then make additionally with the best-performing parameters from the tuning run a new unbiased run on the test data.

Usage

```
tdmEnvTSensi(envT, ind)
```


Arguments

envT results from a prior tuning run. If envT\$tdm\$run==0
 ind an integer from 1:length(envT\$bstGrid): Take the tuning run with index ind.

Value

finals, either NULL (no unbiased run) or a one-row data frame with the results from the unbiased run

Examples

```
## The best results are read from demo02sonar/demoSonar.RData relative to the TDMR
## package directory.
oldwd <- getwd(); setwd(paste(find.package("TDMR"), "demo02sonar", sep="/"));
envT = tdmEnvTLoad("demoSonar.RData"); # loads envT
source("main_sonar.r");
envT$tdm$nrnrun=0; # =0: don't, >0: do unbiasedRun with opts$NRUN=envT$tdm$nrnrun
finals = tdmEnvTSensi(envT,1);
if (!is.null(finals)) print(finals);
setwd(oldwd);
```

tdmExecSpotStep	<i>Execute the tuning process with the given spotStep</i>
-----------------	---

Description

If spotStep=="auto", construct envT from the general TDMR settings in tdm.
 If spotStep=="rep" or "report", load envT from file tdm\$filenameEnvT (from a prior tuning run).
 Then call [tdmBigLoop](#)(envT, spotStep).

Usage

```
tdmExecSpotStep(tdm, spotStep)
```

Arguments

tdm a list with general settings for TDMR, see [tdmDefaultsFill](#)
 spotStep a string, either "auto", "rep" or "report"

Details

If tdm\$filenameEnvT is NULL, set it to sub`(".conf", ".RData", tdm$runList[1])`.

Value

Environment envT containing (among others) the elements

runList	=tdm\$runList
spotList	=tdm\$spotList
tdm	=tdm
getBst	accessor function(confFile,nExp,theTuner) into envT\$bstGrid
getRes	accessor function(confFile,nExp,theTuner) into envT\$resGrid
sCList	list of spotConfig objects, as many as envT\$runList has elements. Each spot-Config object sCList[[k]] contains a list opts as element, which is read from .apd file specified in envT\$runList[k].

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>)

See Also

[tdmBigLoop](#), [tdmEnvTMakeNew](#)

tdmGraAndLogFinalize *Finalize graphics and log file*

Description

Finalize graphics and log file

Usage

```
tdmGraAndLogFinalize(opts, gdObj = NULL)
```

Arguments

opts	with opts\$GD.DEVICE one out of ["pdf" "png" "win" "rstudio" "non"], see tdmGraphicInit
gdObj	object of class TDMgdev, the return value from tdmGraAndLogInitialize, to ensure that tdmGraAndLogInitialize was called before (and the sink on opts\$LOGFILE can be closed)

Initialize graphics and log file.

Description

The log file is opened in `opts$dir.output/opts$LOGFILE`, but only if `opts$fileMode==TRUE`.

Usage

```
tdmGraAndLogInitialize(opts)
```

Arguments

`opts` with `opts$GD.DEVICE` one out of ["pdf" | "png" | "win" | "rstudio" | "non"], see [tdmGraphicInit](#)

Value

`gdObj`, an object of class `TDMgdev`. Pass this object on when calling `tdmGraAndLogFinalize(opts,gdObj)` (if not, a warning is issued before the sink-closing-error occurs)

Close all open graphic devices.

Description

Close all open graphic devices.

Usage

```
tdmGraphicCloseDev(opts, ...)
```

Arguments

`opts` with `opts$GD.DEVICE` one out of ["pdf" | "png" | "win" | "rstudio" | "non"], see [tdmGraphicInit](#)

`...` optional arguments (currently not used)

tdmGraphicCloseWin *Close active file ("png").*

Description

Close active file ("png").

Usage

tdmGraphicCloseWin(opts, ...)

Arguments

opts with opts\$GD.DEVICE one out of ["pdf" | "png" | "win" | "rstudio" | "non"], see [tdmGraphicInit](#)

... optional arguments (currently not used)

tdmGraphicInit *Initialize graphic device.*

Description

Open multipage PDF or (create and) clear opts\$GD.PNGDIR.

Usage

tdmGraphicInit(opts, ...)

Arguments

opts with opts\$GD.DEVICE one out of ["pdf" | "png" | "win" | "rstudio" | "non"]

"pdf" plot everything in one multipage pdf file opts\$PDFFILE

"png" each plot goes into a new png file in opts\$GD.PNGDIR

"win" each plot goes into a new window (dev.new())

"rstudio" plot everything to the RStudio plot device (has a history)

"non" all plots are suppressed

... optional arguments to hand over to [pdf](#) (the other devices require no further arguments)

tdmGraphicNewWin	<i>Initialize a new window.</i>
------------------	---------------------------------

Description

Initialize a new window ("win") / a new file ("png") for current graphic device.

Usage

```
tdmGraphicNewWin(opts, ...)
```

Arguments

opts	with opts\$GD.DEVICE one out of ["pdf" "png" "win" "rstudio" "non"], see tdmGraphicInit
...	optional arguments to hand over to png or windows or X11 in package grDevices (the other devices require no further arguments)

tdmGraphicToTop	<i>Bring the active window to the top</i>
-----------------	---

Description

Only relevant for opts\$GD.DEVICE=="win".

Usage

```
tdmGraphicToTop(opts)
```

Arguments

opts	with opts\$GD.DEVICE one out of ["pdf" "png" "win" "rstudio" "non"], see tdmGraphicInit
------	---

tdmMapDesApply	<i>Apply the mapping from des to opts.</i>
----------------	--

Description

For each variable which appears in .roi (and thus in .des file and design point data frame des): set its counterpart in list opts to the values of the k-th row in des. For each variable not appearing: leave its counterpart in opts at its default value from .apd file.

Usage

```
tdmMapDesApply(des, opts, k, spotConfig, tdm)
```

Arguments

des	design points data frame
opts	list of options
k	apply mapping for the k-th design point
spotConfig	list, we needed here spotConfig\$alg.roi and envT\$mapUser, see tdmMapDesLoad , and in addition envT\$spotConfig\$alg.roi
tdm	list, we need here tdm\$map and tdm\$mapUser

Value

opts, the modified list of options

See Also

[tdmMapDesLoad](#)

tdmMapDesLoad	<i>Load the mapping files.</i>
---------------	--------------------------------

Description

Load the map files "tdmMapDesign.csv" and optionally also "userMapDesign.csv" and store them in tdm\$map and tdm\$mapUser, resp. These maps are used by [tdmMapDesApply](#). "tdmMapDesign.csv" is searched in the TDMR library path find.package("TDMR"). (For the developer version: <tdm\$tdmPath>/inst). "userMapDesign.csv" is searched in tdm\$path (which is getwd() if the user did not define tdm\$path).

Usage

```
tdmMapDesLoad(tdm = list())
```

Arguments

tdm list, needed for tdm\$tdmPath and tdm\$path

Value

tdm, the modified list with new elements tdm\$map and tdm\$mapUser

See Also

[tdmMapDesApply](#)

tdmModConfmat	<i>Calculate confusion matrix, gain and RGain measure.</i>
---------------	--

Description

Calculate confusion matrix, gain and RGain measure.

Usage

```
tdmModConfmat(d, colreal, colpred, opts, predProb = NULL)
```

Arguments

d	data frame
colreal	name of column in d which contains the real class
colpred	name of column in d which contains the predicted class
opts	a list from which we use the elements: <ul style="list-style-type: none"> • gainmat: the gain matrix for each possible outcome, same size as cm\$mat (see below). gainmat[R1,P2] is the gain associated with a record of real class R1 which we predict as class P2. (gain matrix = - cost matrix) • rgain.type: one out of {"rgain" "meanCA" "minCA" "arROC" "arLIFT" "arPRE" }, affects output cm\$mat and cm\$rgain, see below.
predProb	if not NULL, a data frame with as many rows as data frame d, containing columns (index, true label, predicted label, prediction score). Is only needed for opts\$rgain.type=="ar*".

Value

cm, a list containing:

mat	matrix with real class levels as rows, predicted class levels columns. mat[R1,P2] is the number of records with real class R1 predicted as class P2, if opts\$rgain.type=="rgain". If opts\$rgain.type=="meanCA" or "minCA", then show this number as percentage of "records with real class R1" (percentage of each row). CAUTION: If colpred contains NA's, those cases are missing in mat (!) (but the class errors are correct as long as there are no NA's in colreal)
cerr	class error rates, vector of size nlevels(colreal)+1. cerr[X] is the misclassification rate for real class X. cerr["Total"] is the total classification error rate.
gain	the total gain (sum of pointwise product opts\$gainmat*cm\$mat)
gain.vector	gain.vector[X] is the gain attributed to real class label X. gain.vector["Total"] is again the total gain.
gainmax	the maximum achievable gain, assuming perfect prediction
rgain	ratio gain/gainmax in percent, if opts\$rgain.type=="rgain"; mean class accuracy percentage (i.e. mean(diag(cm\$mat))), if opts\$rgain.type=="meanCA"; min class accuracy percentage (i.e. min(diag(cm\$mat))), if opts\$rgain.type=="minCA"; area under ROC curve (a number in [0,1]), if opts\$rgain.type=="arROC"; area between lift curve and horizontal line 1.0, if opts\$rgain.type=="arLIFT"; area under precision-recall curve (a number in [0,1]), if opts\$rgain.type=="arPRE";

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), Patrick Koch

See Also

[tdmClassify](#) [tdmROCRbase](#)

tdmModCreateCVindex *Create and return a training-validation-set index vector.*

Description

Depending on the value of member TST.kind in list opts, the returned index cvi is

1. TST.kind="cv": a random cross validation index P([111...222...333...]) - or -
2. TST.kind="rand": a random index with P([00...11...-1-1...]) for training (0), validation (1) and disregard (-1) cases - or -
3. TST.kind="col": the column dset[opts\$TST.COL] contains the training (0), validation (1) and disregard (-1) set division (and all records with a value <0 in column TST.COL are disregarded).

Here $P(\cdot)$ denotes random permutation of the sequence.

The disregard set is optional, i.e. `cvi` may contain only 0 and 1, if desired.

Special case `TST.kind="cv"` and `TST.NFOLD=1`: make *every* record a training record, i.e. index [000...].

In case `TST.kind="rand"` and `stratified=TRUE` a *stratified* sample is drawn, where the strata in the training case reflect the rel. frequency of each level of the *1st* response variable and are ensured to be at least of size 1.

Usage

```
tdmModCreateCVindex(dset, response.variables, opts, stratified = FALSE)
```

Arguments

<code>dset</code>	the data frame for which <code>cvi</code> is needed
<code>response.variables</code>	issue a warning if <code>length(response.variables)>1</code> . Use the first response variable for determining strata size.
<code>opts</code>	a list from which we need here the following entries <ul style="list-style-type: none"> • <code>TST.kind</code>: ["cv" "rand" "col"] • <code>TST.NFOLD</code>: number of CV folds (only relevant in case <code>TST.kind=="cv"</code>) • <code>TST.COL</code>: column of <code>dset</code> containing the (0/1/<0) index (only relevant in case <code>TST.kind=="col"</code>) or NULL if no such column exists • <code>TST.valiFrac</code>: fraction of records to set aside for validation (only relevant in case <code>TST.kind=="rand"</code>) • <code>TST.trnFrac</code>: [1-opts\$TST.valiFrac] fraction of records to use for training (only relevant in case <code>TST.kind=="rand"</code>)
<code>stratified</code>	[F] do stratified sampling for <code>TST.kind="rand"</code> with at least one training record for each response variable level (classification)

Value

`cvi` training-validation-set (0/>0) index vector (all records with `cvi<0`, e.g. from column `TST.COL`, are disregarded)

Note

Currently stratified sampling in case `TST.KIND='rand'` does only work correctly for *one* response variable. If there are more than one, the right fraction of validation records is taken, but the strata are drawn w.r.t. the first response variable. (For multiple response variables we would have to return a list of `cvi`'s or to call `tdmModCreateCVindex` for each response variable anew.)

tdmModSortedRFimport *Sort the input variables decreasingly by their RF-importance.*

Description

Build a Random Forest using importance=TRUE. Usually the RF is smaller (50 trees), to speed up computation. Use na.roughfix for missing value replacement. Decide which input variables to keep and return them in SRF\$input.variables

Usage

```
tdmModSortedRFimport(d_train, response.variable, input.variables, opts)
```

Arguments

d_train	training set
response.variable	the target column from d_train to use for the RF-model
input.variables	the input columns from d_train to use for the RF-model
opts	options, here we use the elements [defaults in brackets]:

- SRF.kind:
 - = "xperc": keep a certain importance percentage, starting from the most important variable
 - = "ndrop": drop a certain number of least important variables
 - = "nkeep": keep a certain number of most important variables
 - = "none": do not call `tdmModSortedRFimport` at all (see `tdmRegress.r` and `tdmClassify.r`)
- SRF.ndrop: [0] how many variables to drop (if SRF.kind=="ndrop")
- SRF.XPerc: [0.95] if >=0, keep that importance percentage, starting with the most important variables (if SRF.kind=="xperc")
- SRF.calc: [TRUE] =TRUE: calculate importance & save on SRF.file, =F: load from SRF.file (SRF.file = Output/<filename>.SRF.<response.variable>.Rdata)
- SRF.ntree: [50] number of RF trees
- SRF.verbose: [2]
- SRF.maxS: [40] how many variables to show in plot
- SRF.minlsi: [1] a lower bound for the length of SRF\$input.variables
- RF.sampsize: sampsize for RF, set prior to calling this func via `tdmModAdjustSampsize(opts$SRF.samp,...)`
- GD.DEVICE: if != "non", then make a bar plot on current graphic device
- CLS.CLASSWT: class weight vector to use in random forest training

Value

SRF, a list with the following elements:

<code>input.variables</code>	the vector of input variables which remain after importance processing. These are sorted by decreasing importance.
<code>s_input</code>	all <code>input.variables</code> sorted by decreasing (**NEW**) importance
<code>s_imp1</code>	the importance for <code>s_input</code>
<code>s_dropped</code>	vector with name of dropped variables
<code>lsd</code>	length of <code>s_dropped</code>
<code>perc</code>	the percentage of total importance which is in the dropped variables
<code>opts</code>	some defaults might have been added

Author(s)

Wolfgang Konen, Patrick Koch <wolfgang.konen@fh-koeln.de>

tdmModVote2Target *Analyze how the vote fraction corresponds to reliability of prediction.*

Description

This function analyzes whether in different vote bins the trained RF makes predictions with different reliability. Only for RF-prediction in case of binary (0/1) classification.

Expected result: The larger the fraction of trees voting for class 0 is, the smaller is the percentage of true class-1- cases in this vote bin. This function is somewhat specialized for the DMC2010-task.

Usage

```
tdmModVote2Target(vote0, pred, target)
```

Arguments

<code>vote0</code>	vector: which fraction of trees votes for class 0?
<code>pred</code>	vector: the predicted class for each record (0/1)
<code>target</code>	vector: the true class for each vector (0/1)

Value

a data frame with columns

<code>vcut</code>	vote cut v
<code>count</code>	number of cases with vote fraction in $[v[i-1],v[i]]$
<code>pred0</code>	fraction of 0-predictions
<code>pCorr</code>	fraction of correct predictions
<code>pR</code>	fraction of true 1-cases

Author(s)

Wolfgang Konen <wolfgang.konen@fh-koeln.de>

tdmOptsDefaultsSet *Default values for list opts.*

Description

Set up and return a list `opts` with default settings. The list `opts` contains all DM-related settings which are needed by `main_<TASK>`.

For better readability, most elements of `opts` are arranged in groups:

<code>dir.*</code>	path-related settings
<code>READ.*</code>	data-reading-related settings
<code>TST.*</code>	resampling-related settings (training, validation and test set, CV)
<code>PRE.*</code>	preprocessing parameters
<code>SRF.*</code>	several parameters for <code>tdmModSortedRFimport</code>
<code>MOD.*</code>	general settings for models and model building
<code>RF.*</code>	several parameters for model RF (Random Forest)
<code>SVM.*</code>	several parameters for model SVM (Support Vector Machines)
<code>ADA.*</code>	several parameters for model ADA (AdaBoost)
<code>CLS.*</code>	classification-related settings
<code>GD.*</code>	settings for the graphic devices

Usage

```
tdmOptsDefaultsSet(opts = NULL, path = "./")
```

Arguments

<code>opts</code>	(optional) the options already set
<code>path</code>	["./"] the path string to precede all settings <code>opts\$dir.*</code> . If <code>path=="./"</code> then nothing is preceded to <code>opts\$dir.*</code> .

Details

The path-related settings are relative to `dir(tdm$mainFile)`, if it is def'd, else relative to the current dir.

Finally, the function `tdmOptsDefaultsFill(opts)` is called to fill in further details, depending on the current settings of `opts`.

Value

a list `opts`, with defaults set for all options relevant for a DM task, containing the following elements

dir.txt	[<path>/data] where to find .txt/.csv files
dir.data	[<path>/data] where to find other data files, including .Rdata
dir.Rdata	[<path>/Rdata] – deprecated, use opts\$dir.data –
dir.output	[<path>/Output] where to put output files
filename	["default.txt"] the task data
filetest	[NULL] the test data, only relevant for READ.TST=T
fileMode	[TRUE] if =T, write opts\$EVALFILE=*_train_eval.csv, *_train.csv.SRF.*.RData file and *_train.log file
logFile	[TRUE] if =T, and if opts\$fileMode=T, write log file to *_train.log
data.title	["Default Data"] title for plots
READ.TXT	[T] =T: read data from .csv and save as .Rdata, =F: read from .Rdata
READ.NROW	[-1] read this amount of rows or -1 for 'read all rows'
READ.TST	[F] =T: read unseen test data from opts\$filetest (usually you will do this only for the final model and only with TST.kind="col")
READ.CMD	[" tdmReadCmd (filename, opts)"] the command to be passed into tdmReadData . It has to contain the placeholder 'filename'. The default in brackets implies 'read.csv(file=paste(opts\$dir.txt, filename, sep=""), nrow=opts\$READ.NROW)' which includes the further settings header=T, sep="," and dec="."
READ.INI	[TRUE] read the task data initially, i.e. prior to tuning, using tdmReadData . If =FALSE, the data are read anew in each pass through main_TASK, i.e. in each tuning step (deprecated).
TST.kind	["rand"] one of the choices from {"cv","rand","col"}, see tdmModCreateCVindex for details
TST.COL	["TST.COL"] name of column with train/test/disregard-flag
TST.NFOLD	[3] number of CV-folds (only for TST.kind=="cv")
TST.valiFrac	[0.1] set this fraction of the train-validation data aside for validation (only for TST.kind=="rand")
TST.testFrac	[0.1] set prior to tuning this fraction of data aside for testing (if tdm\$umode=="SP_T" and opts\$READ.INI==TRUE) or set this fraction of data aside for testing after tuning (if tdm\$umode=="RSUB" or == "CV")
TST.trnFrac	[NULL] train set fraction, if NULL then tdmModCreateCVindex will set it to 1 - opts\$TST.valiFrac.
TST.SEED	[NULL] a seed for the random test set selection (tdmRandomSeed) and random validation set selection. (tdmClassifyLoop). If NULL, use tdmRandomSeed .
PRE.PCA	["none" (default) "linear"] PCA preprocessing: [don't do normal PCA (prcomp)]
PRE.PCA.REPLACE	[T] =T: replace with the PCA columns the original numerical columns, =F: add the PCA columns
PRE.PCA.npc	[0] if >0: add monomials of degree 2 from the first PRE.PCA.npc columns (PCs) (only active, if opts\$PRE.PCA!="none")

PRE.SFA	["none" (default)!"2nd"] SFA preprocessing (see package rSFA : [don't do ormal SFA with 2nd degree expansion]
PRE.SFA.REPLACE	[F] =T: replace the original numerical columns with the SFA columns; =F: add the SFA columns
PRE.SFA.npc	[0] if >0: add monomials of degree 2 from the first PRE.SFA.npc columns (only active, if opts\$PRE.SFA!="none")
PRE.SFA.PPRANGE	[11] number of inputs after SFA preprocessing, only those inputs enter into SFA expansion
PRE.SFA.ODIM	[5] number of SFA output dimensions (slowest signals) to return
PRE.SFA.doPB	[T] =F T: don't do parametric bootstrap for SFA in case of marginal training data
PRE.SFA.fctPB	[sfaPBootstrap] the function to call in case of parametric bootstrap, see sfaPBootstrap in package rSFA for its interface description
PRE.allNonValid	[F] if =T, then use all non-validation data in the training-validation set for PCA or SFA preprocessing. If =F, use only the training set for PCA or SFA processing (only relevant if opts\$PRE.PCA!="none" or opts\$PRE.SFA!="none").
PRE.Xpgroup	[0.99] bind the fraction 1-PRE.Xpgroup in column OTHER (see tdmPreGroupLevels)
PRE.MaxLevel	[32] bind the N-32+1 least frequent cases in column OTHER (see tdmPreGroupLevels)
SRF.kind	["xperc" (default)!"ndrop"!"nkeep"!"none"] the method used for feature selection, see tdmModSortedRFimport
SRF.ndrop	[0] how many variables to drop (only relevant if SRF.kind=="ndrop")
SRF.nkeep	[NULL] how many variables to keep, NULL="keep all" (only relevant if SRF.kind=="nkeep")
SRF.XPerc	[0.95] if >=0, keep that importance percentage, starting with the most important variables (if SRF.kind=="xperc")
SRF.calc	[T] =T: calculate importance & save on SRF.file, =F: load from srfFile (srfFile = Output/<confFile>.SRF.Rdata)
SRF.ntree	[50] number of RF trees
SRF.samp	sampsize for RF in importance estimation. See RF.samp for further info on sampsize.
SRF.verbose	[2]
SRF.maxS	[40] how many variables to show in plot
SRF.minlsi	[1] a lower bound for the length of SRF\$input.variables
SRF.method	["RFimp"]
SRF.scale	[TRUE] option 'scale' for call importance() in tdmModSortedRFimport
MOD.SEED	[NULL] a seed for the random model initialization (if model is non-deterministic). If NULL, use tdmRandomSeed .
MOD.method	["RF" (default)!"MC.RF"!"SVM"!"NB"]: use [RF MetaCost-RF SVM Naive Bayes] in tdmClassify ["RF" (default)!"SVM"!"LM"]: use [RF SVM linear model] in tdmRegress

RF.ntree	[500]
RF.samp	[NULL] sampsize for RF in model training. If RF.samp is a scalar, then it specifies the total size of the sample. For classification, it can also be a vector of length n.class (= # of levels in response variable), then it specifies the size of each strata. The sum of the vector is the total sample size. If NULL, RF.samp will be replaced by 3000 later in tdmModAdjustSampsize*.
RF.mtry	[NULL]
RF.nodesize	[1]
RF.OOB	[TRUE] if =T, return OOB-training set error as tuning measure; if =F, return validation set error
RF.p.all	[FALSE]
SVM.kernel	[3] =1: linear, =2: polynomial, =3: RBF, =4: sigmoid
SVM.epsilon	[0.005] needed only for regression
SVM.gamma	[0.005]
SVM.coef0	[0.0] (needed only for opts\$SVM.kernel=="polynomial" or == "sigmoid")
SVM.degree	[3] (needed only for opts\$SVM.kernel=="polynomial")
SVM.tolerance	[0.008]
ADA.coeflearn	[1]=1: "Breiman", =2: "Freund", =3: "Zhu" as value for boosting(...,coeflearn,...) (AdaBoost)
ADA.mfinal	[10] number of trees in AdaBoost = mfinal boosting(...,mfinal,...)
ADA.rpart.minsplit	[20] minimum number of observations in a node in order for a split to be attempted
CLS.cutoff	[NULL] vote fractions for the classes (vector of length n.class = # of levels in response variable). The class i with maximum ratio (% votes)/CLS.cutoff[i] wins. If NULL, then each class gets the cutoff 1/n.class (i.e. majority vote wins). The smaller CLS.cutoff[i], the more likely class i will win.
CLS.CLASSWT	[NULL] class weights for the n.class classes, e.g. c(A=10,B=20) for a 2-class problem with classes A and B (the higher, the more costly is a misclassification of that real class). It should be a named vector with the same length and names as the levels of the response variable. If no names are given, the levels of the response variables in lexicographical order will be attached in <code>tdmClassify</code> . CLS.CLASSWT=NULL for no weights.
CLS.gainmat	[NULL] (n.class x n.class) gain matrix. If NULL, CLS.gainmat will be set to unit matrix in <code>tdmClassify</code>
rgain.type	["rgain" (default) "meanCA" "minCA"] in case of <code>tdmClassify</code> : For classification, the measure Rgain returned from <code>tdmClassifyLoop</code> in <code>result\$R_*</code> is [relative gain (i.e. gain/gainmax) mean class accuracy minimum class accuracy]. The goal is to maximize Rgain. For binary classification there are the additional measures ["arROC" "arLIFT" "arPRE"], see <code>tdmModConfmat</code> . For regression, the goal is to minimize <code>result\$R_*</code> returned from <code>tdmRegress</code> .

	In this case, possible values are <code>rgain.type = ["rmae" (default) "rmse" "made"]</code> which stands for [relative mean absolute error root mean squared error mean absolute deviation].
<code>ncopies</code>	[0] if >0, activate tdmParaBootstrap in tdmClassify
<code>fct.postproc</code>	[NULL] name of a function with signature <code>(pred, dframe, opts)</code> where <code>pred</code> is the prediction of the model on the data frame <code>dframe</code> and <code>opts</code> is this list. This function may do some postprocessing on <code>pred</code> and it returns a (potentially modified) <code>pred</code> . This function will be called in tdmClassify if it is not NULL.
<code>GD.DEVICE</code>	["win"] = "win": all graphics to (several) windows (windows or X11 in package <code>grDevices</code>) ="rstudio": same as "win", but all graphics go to the RStudio device ="pdf": all graphics to one multi-page PDF ="png": all graphics in separate PNG files in <code>opts\$GD.PNGDIR</code> ="non": no graphics at all This concerns the TDMR graphics, not the SPOT (or other tuner) graphics. If running R from RStudio (if there is a device with name "RStudioGD") then the default "win" is changed to "rstudio" automatically.
<code>GD.RESTART</code>	[T] =T: restart the graphics device (i.e. close all 'old' windows or re-open multi-page pdf) in each call to tdmClassify or tdmRegress , resp. =F: leave all windows open (suitable for calls from SPOT) or write more pages in same pdf.
<code>GD.CLOSE</code>	[T] =T: close graphics device "png", "pdf" at the end of <code>main_*.r</code> (suitable for <code>main_*.r</code> solo) or =F: do not close (suitable for call from <code>tdmStartSpot</code> , where all windows should remain open)
<code>NRUN</code>	[2] how many runs with different train & test samples - or - how many CV-runs, if <code>opts\$TST.kind="cv"</code>
<code>APPLY_TIME</code>	[FALSE]
<code>test2.show</code>	[FALSE]
<code>test2.string</code>	["default cutoff"]
<code>VERBOSE</code>	[2] =2: print much output, =1: less, =0: none

Note

The variables `opts$PRE.PCA.numericV` and `opts$PRE.SFA.numericV` (string vectors of numeric input columns to be used for PCA or SFA) are not set by [tdmOptsDefaultsSet](#) or [tdmOptsDefaultsFill](#). Either they are supplied by the user or, if NULL, TDMR will set them to `input.variables` in [tdmClassifyLoop](#), assuming that all columns are numeric.

Author(s)

Wolfgang Konen, FHK, 2011 - 2013

See Also

[tdmOptsDefaultsFill](#) [tdmDefaultsFill](#)

tdmParaBootstrap	<i>Parametric bootstrap: add 'noisy copies' to a data frame (training data).</i>
------------------	--

Description

A normal distribution is approximated from the data given in `dset[,input.variables]` and new data are drawn from this distribution for the columns `input.variables`. The column `resp` is filled at random with levels with the same relative frequency as in `dset[,resp]`. Other columns of `dset` are filled by copying the entries from the first row of `dset`.

Usage

```
tdmParaBootstrap(dset, resp, input.variables, opts)
```

Arguments

<code>dset</code>	data frame with training set
<code>resp</code>	name of column in <code>dset</code> which carries the target variable
<code>input.variables</code>	vector with names of input columns
<code>opts</code>	additional parameters [defaults in brackets]
	<code>ncopies</code> how many noisy copies to add
	<code>ncsigma</code> [1.0] multiplicative factor for each <code>std.dev.</code>
	<code>ncmethod</code> [3] which method to use for parametric bootstrap
	=1: each 'old' record from <code>X</code> in turn is the centroid for a new pattern;
	=2: the centroid is the average of all records from the same class, the <code>std.dev.</code> is the same for all classes;
	=3: centroid as in '2', the <code>std.dev.</code> is the <code>std.dev.</code> of all records from the same class (*recommended*)
	<code>TST.COL</code> (optional) name of column in <code>dset</code> where each PB record is marked with a 0

Value

data frame `dset` with the new parametric bootstrap records added as last rows.

Author(s)

Wolfgang Konen, FHK, Nov'2011-Dec'2011

See Also

[tdmClassify](#)

tdmPlotResMeta

*Interactive plots of RES data frames and their metamodels.***Description**

Makes interactive plots for any of the result data frames contained in `envT` together with fitted metamodels. `tdmPlotResMeta` creates a [twiddle](#) interface which allows to select

- `tuner`: one of the tuners
- `nExperim`: a knob to select one of the experiments in `envT` (only if `envTtdmnExperim > 1`)
- `reportFunc`: one of {"spotReport3d", "spotReportContour"}, see [SPOT](#)
- `modelFit`: one of {"spotPredictGausspr", "spotPredictRandomForest", "spotPredictForrester", "spotPredictMleqp"}, see [SPOT](#)
- `skipLastSteps`: [0] skip the skipLastSteps sequential steps from the selected RES data frame.
- `skipIncomplete`: [FALSE] if this checkbox is checked (i.e. `skipIncomplete==TRUE`), then all CONFIGs having *less* than `seq.design.maxRepeats` repeats are skipped
- `nSkip`: [0] skip the nSkip worst CONFIGs
- `y10Exp`: [0] multiply in the RES data frame the column Y (objective function) by the factor 10^{y10Exp} . This affects the coloring of the metamodel in case "spotReport3d": A surface in the Y-range [1e-3, 1e-2] would have only a single color, but a surface in the range [10,100] will have a richer color scheme.

Usage

```
tdmPlotResMeta(envT)
```

Arguments

`envT` environment with results as returned from [tdmBigLoop](#) or as loaded from an appropriate `.RData` file. We use here especially `envT$resGrid`, `envT$tdm` and `envT$tunerVal` (the `spotConfig` from the last tuning experiment).

Details

In case of "spotReport3d", the plot shows the metamodel as colored surface and the true observations of the objective function as black points. Each point stands for a CONFIG, i.e. it is an aggregation (usually: mean) of all repeats for a design point configuration. The options `nSkip` and `skipIncomplete` allow to suppress certain CONFIGs which may be outliers and thus hinder the view on the 'interesting' region. Note however, that this suppression also affects the metamodel fit, so use these options with care.

In case of "spotReportContour", the plot shows a contour plot of the metamodel with a black point indicating the best CONFIG found.

If `envT$tunerVal$meta.compare` is `TRUE`, the quality of the metamodel is evaluated with the following procedure: A RES data frame `newRes` with `N` new design CONFIGs (not used during metamodel fit) is taken and their true target value (evaluation of target function) is compared with

the metamodel's value. The mean absolute deviation

```
sum(|metamodel_i - newCONFIG_i|) / N
```

is printed and returned in `envT$tunerVal$meta.newMAD`.

The RES data frame `newRes` is either taken from `envT$tunerVal$meta.newRes` (if this element is not NULL) or it is taken from another member of `envT$resGrid` (preferably from a 'lhd' tuner result, because it is evenly distributed).

If `envT$tunerVal$meta.compare` is NULL, it is set to FALSE.

Note

Side Effects: In case of "spotReport3d", one or several RGL plot windows are created which can be manipulated interactively. A certain RGL window `n` can be selected with `rgl.set(n)`. An interactively manipulated RGL window can be saved with `rgl.snapshot("myplot.png")`.

In case of "spotPredictMlegp" there is a new element `sC$seq.mlegp.min.nugget=1` set in the source code which causes the MLEGP fit to become much smoother in the presence of a noisy target function (nugget effect).

Author(s)

Wolfgang Konen

See Also

[tdmBigLoop](#)

Examples

```
## Not run:
##
## Read previous tuning results 'envT' from demo02sonar/demoSonar.RData
## (relative to the TDMR package directory).
## Then, tdmPlotResMeta lets you explore interactively the RES data frame(s):
load(paste(find.package("TDMR"), "demo02sonar", "demoSonar.RData", sep="/"));
tdmPlotResMeta(envT);

## End(Not run)
```

`tdmPreAddMonomials` *Add monomials of degree 2 to a data frame.*

Description

Given the data frame `dset` and a data frame `rx` with the same number of rows, add monomials of degree 2 to `dset` for all quadratic combinations of the first `PRE.npc` columns of `rx`. The naming of these new columns is "R1x2" for the combination of cols 1 and 2 and so on (if `prefix="R"`).

Usage

```
tdmPreAddMonomials(dset, rx, PRE.npc, opts, degree = 2, prefix = "R")
```

Arguments

dset	the target data frame
rx	a data frame where to draw the monomials from
PRE.npc	the number of columns from rx to use (clipped to ncol(rx) if necessary)
opts	a list from which we need here the following entries: <ul style="list-style-type: none"> • filename • VERBOSE
degree	[2] (currently only 2 is supported)
prefix	["R"] character prefix for the monomial column names

Value

data frame dset with the new monomial columns appended. If PRE.npc==0, the data frame is returned unchanged.

Note

CAVEAT: The double for-loop costs some time (e.g. 2-4 sec for ncol(rx)=8 or 10) How to fix: make a version w/o for-loop and w/o frequent assigns to dset (**TODO**)

tdmPreFindConstVar *Find constant columns.*

Description

Find all those columns in data frame dset which are completely constant or completely NA and return a vector with their names.

Usage

```
tdmPreFindConstVar(dset)
```

Arguments

dset	data frame
------	------------

Value

name vector of constant columns

tdmPreGroupLevels *Group the levels of factor variable in dset[, colname].*

Description

This function reduces the number of levels for factor variables with too many levels. It counts the cases in each level and orders them decreasingly. It binds the least frequent levels together in a new level "OTHER" such that the remaining untouched levels have more than `opts$PRE.Xpgroup` percent of all cases. OR it binds the levels with least cases together in "OTHER" such that the total number of new levels is `opts$PRE.MaxLevel`. From these two choices for "OTHER" take the one which binds more variables in column "OTHER".

Usage

```
tdmPreGroupLevels(dset, colname, opts)
```

Arguments

dset	data frame
colname	name of column to be re-grouped
opts	list, here we need <ul style="list-style-type: none"> • <code>PRE.Xpgroup</code> [0.99] • <code>PRE.MaxLevel</code> [32] (32 is the maximum number of levels allowed for randomForest)

Value

dset, a data frame with `dset[, colname]` re-grouped

tdmPreLevel2Target *Relate levels of a column with a target (column).*

Description

Print for each level of factor variable `f` which ratio 0 / 1 of the binary target variable it contains and how many cases are in each level

Usage

```
tdmPreLevel2Target(dset, target, f, opts)
```

Arguments

dset	data frame
target	name of target column
f	number of column with factor variable
opts	list, here we need <ul style="list-style-type: none">• opts\$thresh_pR• opts\$verbose

Note

SIDE EFFECTS: some printed output

tdmPreNAroughfix *Replace <NA> values with suitable non <NA> values*

Description

This function replaces <NA> values in a list entry or data frame column with the median (for numeric columns) or the most frequent mode (for factor columns). It does the same as `na.roughfix` in package `randomForest`, but does so faster.

Usage

```
tdmPreNAroughfix(object, ...)
```

Arguments

object	list or data frame
...	additional arguments

Value

object, the list or data frame with <NA> values replaced

tdmPrePCA.apply	<i>Apply PCA (Principal Component Analysis) to new data.</i>
-----------------	--

Description

The PCA rotation is taken from `pcaList`, a value returned from a prior call to `tdmPrePCA.train`.

Usage

```
tdmPrePCA.apply(dset, pcaList, opts, dtrain = NULL)
```

Arguments

<code>dset</code>	the data frame with the new data
<code>pcaList</code>	a value returned from a prior call to <code>tdmPrePCA.train</code>
<code>opts</code>	a list from which we need here the following entries: <ul style="list-style-type: none"> • <code>PRE.knum</code>: if >0 and if <code>PRE.PCA="kernel"</code>, take only a subset of <code>PRE.knum</code> records from <code>dset</code> • <code>PRE.PCA.npc</code>: if >0, then add for the first <code>PRE.PCA.npc</code> PCs the monomials of degree 2 (see <code>tdmPreAddMonomials</code>) • <code>PRE.PCA.numericV</code> vector with all column names in <code>dset</code> for which PCA is performed. These columns may contain <i>numeric</i> values only.
<code>dtrain</code>	[NULL] optional, only needed in case that <code>dset</code> is a 0-row-data frame: then we 'borrow' the columns from <code>dtrain</code> , the data set returned from <code>tdmPrePCA.train</code> in <code>pca\$dset</code> .

Value

`pca`, a list with entries:

<code>dset</code>	the input data frame <code>dset</code> with columns <code>numeric.variables</code> replaced by the PCs with names <code>PC1</code> , <code>PC2</code> , ... (in case <code>PRE.PCA=="linear"</code>) or with names <code>KP1</code> , <code>KP2</code> , ... (in case <code>PRE.PCA=="kernel"</code>) and optional with monomial columns added, if <code>PRE.PCA.npc>0</code>
<code>numeric.variables</code>	the new column names for PCs and for the monomials

Author(s)

Wolfgang Konen, FHK, Mar'2011 - Jan'2012

See Also

[tdmPrePCA.train](#)

tdmPrePCA.train	<i>PCA (Principal Component Analysis) for numeric columns in a data frame.</i>
-----------------	--

Description

tdmPrePCA.train is capable of linear PCA, based on prcomp (which uses SVD), and of kernel PCA (either KPCA, KHA or KFA).

Usage

```
tdmPrePCA.train(dset, opts)
```

Arguments

dset	the data frame with training (and test) data.
opts	a list from which we need here the following entries: <ul style="list-style-type: none"> • PRE.PCA: ["linear" "kernel" "none"] • PRE.knum: if >0 and if PRE.PCA="kernel", take only a subset of PRE.knum records from dset • PRE.PCA.REPLACE: [T] =T: replace the original numerical columns with the PCA columns; =F: add the PCA columns • PRE.PCA.npc: if >0, then add for the first PRE.PCA.npc PCs the monomials of degree 2 (see tdmPreAddMonomials) • PRE.PCA.numericV vector with all column names in dset for which PCA is performed. These columns may contain *numeric* values only.

Value

pca, a list with entries:

dset	the input data frame dset with columns numeric.variables replaced or extended (depending on opts\$PRE.PCA.REPLACE) by the PCs with names PC1, PC2, ... (in case PRE.PCA=="linear") or with names KP1, KP2, ... (in case PRE.PCA=="kernel") and optional with monomial columns added, if PRE.PCA.npc>0. The number of PCs is min(nrows(dset),length(numeric.variables)).
numeric.variables	the new numeric column names (PCs, monomials, and optionally old numericV, if opts\$PRE.PCA.REPLACE==F)
pcaList	a list with the items sdev, rotation, center, scale, x as returned from prcomp plus eival, the eigenvalues for the PCs

Note

CAUTION: Kernel PCA (opts\$PRE.PCA=="kernel") is currently disabled in code, it *crashes* for large number of records or large number of columns.

Author(s)

Wolfgang Konen, FHK, Mar'2011 - Jan'2012

See Also

[tdmPrePCA.apply](#)

tdmPreSFA.apply	<i>Apply SFA (Slow Feature Analysis) to new data.</i>
-----------------	---

Description

The SFA projection is taken from `sfaList`, a value returned from a prior call to `tdmPreSFA.train`.

Usage

```
tdmPreSFA.apply(dset, sfaList, opts, dtrain = NULL)
```

Arguments

<code>dset</code>	the data frame with the new data
<code>sfaList</code>	a value returned from a prior call to <code>tdmPreSFA.train</code>
<code>opts</code>	a list from which we need here the following entries: <ul style="list-style-type: none"> • <code>PRE.SFA.REPLACE</code>: [T]=T: replace the original numerical columns with the SFA columns; =F: add the SFA columns • <code>PRE.SFA.npc</code>: if >0, then add for the first <code>PRE.SFA.npc</code> PCs the monomials of degree 2 (see <code>tdmPreAddMonomials</code>) • <code>PRE.SFA.ODIM</code>: [5] number of SFA output dimensions (slowest signals) to return • <code>PRE.SFA.numericV</code> vector with all column names in <code>dset</code> for which SFA is performed. These columns may contain <code>*numeric*</code> values only.
<code>dtrain</code>	[NULL] optional, only needed in case that <code>dset</code> is a 0-row-data frame: then we 'borrow' the columns from <code>dtrain</code> , the data set returned from <code>tdmPreSFA.train</code> in <code>sfa\$dset</code> .

Value

`sfa`, a list with entries:

<code>dset</code>	the input data frame <code>dset</code> with columns <code>numeric.variables</code> replaced by the PCs with names <code>PC1</code> , <code>PC2</code> , ... (in case <code>PRE.SFA=="linear"</code>) or with names <code>KP1</code> , <code>KP2</code> , ... (in case <code>PRE.SFA=="kernel"</code>) and optional with monomial columns added, if <code>PRE.SFA.npc>0</code>
<code>numeric.variables</code>	the new column names for PCs and for the monomials

Author(s)

Wolfgang Konen, Martin Zaeferrer, FHK, Jan'2012 - Feb'2012

See Also

[tdmPreSFA.train](#)

tdmPreSFA.train	<i>SFA (Slow Feature Analysis) for numeric columns in a data frame.</i>
-----------------	---

Description

tdmPreSFA.train uses package [rSFA](#). It is assumed that classification for the variable contained in column `response.var` is done. SFA seeks features in an expanded function space for which the intra-class variation w.r.t. `response.var` is as low as possible.

Usage

```
tdmPreSFA.train(dset, response.var, opts)
```

Arguments

dset	the data frame with training (and test) data.
response.var	the response variable for classification.
opts	a list from which we need here the following entries: <ul style="list-style-type: none"> • PRE.SFA: ["linear" "2nd" "none"] which stands for [1st 2nd degree monomial SFA no SFA] • PRE.SFA.REPLACE: [T] =T: replace the original numerical columns with the SFA columns; =F: add the SFA columns • PRE.SFA.npc: if >0, then add for the first PRE.SFA.npc PCs the monomials of degree 2 (see <code>tdmPreAddMonomials</code>) • PRE.SFA.PPRANGE: [11] number of inputs after preprocessing, they enter into expansion • PRE.SFA.ODIM: [5] number of SFA output dimensions (slowest signals) to return • PRE.SFA.numericV vector with all column names in dset which are input for SFA. These columns may contain <i>*numeric*</i> values only.

Value

sfa, a list with entries:

dset	the input data frame dset with columns <code>numeric.variables</code> replaced or extended (depending on <code>opts\$PRE.SFA.REPLACE</code>) by the SFA components with names SF1, SF2, ... and with optional monomial columns added, if <code>PRE.SFA.npc>0</code>
------	---

numeric.variables the new numeric column names of dset, i.e. SFA components, monomials (and optionally PRE.SFA.numericV, if opts\$PRE.SFA.REPLACE==F)

sfaList a list with the items opts (sfaOpts), matrices DSF and SF and many others, as returned from [sfaStep](#)

Author(s)

Wolfgang Konen, Martin Zaefferer, FHK, Jan'2012 - Feb'2012

See Also

[tdmPreSFA.apply](#)

tdmRandomSeed	<i>Generates pseudo-random random number seeds.</i>
---------------	---

Description

To use this mechanism, create first an object tdmRandomSeed with a call to [makeTdmRandomSeed](#).

Usage

```
tdmRandomSeed()
```

Value

In each call to this function a different integer in 0...100001+nCall is returned. This is true even if it is called many times within the same second (where Sys.time() will return the same integer). nCall is the number of calls to this function object.

Author(s)

Wolfgang Konen, Patrick Koch <wolfgang.konen@fh-koeln.de>

See Also

[makeTdmRandomSeed](#)

Examples

```
tdmRandomSeed = makeTdmRandomSeed();
for (i in 1:10) print(c(as.integer(Sys.time()), tdmRandomSeed()));
```

tdmReadCmd	<i>Template function, the default for opts\$READ.CMD.</i>
------------	---

Description

This template function is used by [tdmReadData](#) for data reading.

Usage

```
tdmReadCmd(filename, opts)
```

Arguments

filename	this is either opts\$filename or opts\$filetest
opts	list of options, we need here <ul style="list-style-type: none"> • dir.txt: string, the directory where to read filename • READ.NROW: [-1] read only that many rows from opts\$filename. -1 for 'read all rows'.

Value

dset, a data frame with all data read

Note

This default template function will issue the command
`dset=read.csv(file=paste(opts$dir.txt, filename, sep=""),nrow=opts$READ.NROW)`
 which includes the default settings `header=T, sep=","` and `dec="."`. The user may alternatively specify his/her own function, the only rules are

- that it has to return a data frame (dset or tset)
- that the string `opts$READ.CMD` contains an argument filename

([tdmReadData](#) will call this function with either `opts$filename` or `opts$filetest` replacing filename)

tdmReadData	<i>Read the data according to the settings in opts.</i>
-------------	---

Description

Side effect: Data are saved on .Rdata file if `opts$READ.TXT=TRUE` (for quicker next-time access).
 Next-time access: Data are re-loaded from .Rdata if `opts$READ.TXT=FALSE`.

Usage

```
tdmReadData(opts)
```

Arguments

opts	<p>list of options, we need here</p> <ul style="list-style-type: none"> • filename: string of filename in <code>opts\$dir.txt</code> • filetest: string of filename test data in <code>opts\$dir.txt</code> (only if <code>READ.TST=T</code>) • READ.TXT: [T] if =T: read from <code>opts\$filename</code>, =F: load from .Rdata file • READ.CMD: ["tdmReadCmd(filename,opts)"] string with a file-read-command with placeholder 'filename' • READ.TST: [F] if =T: read also test data from <code>filetest</code> • TST.COL: ["TST.COL"] string, create a column with the name of this string in <code>dset</code>, which has 0 for training and 1 for test data • READ.NROW: [-1] read only that many rows from <code>opts\$filename</code>. -1 for 'read all rows'.
------	--

Details

When `opts$READ.TST==T`, the following things happen in `tdmReadData`: Data are read from `opts$filename` and from `opts$filetest`. Both data sets are bound together, with a new column `opts$TST.COL` having '0' for the data from `opts$filename` and having '1' for the data from `opts$filetest`. The branch using `opts$TST.COL` is invoked either with `umode="TST"` in `unbiasedRun` or with `opts$TST.kind="col"` in `tdmModCreateCVindex`.

Value

`dset`, a data frame with all data read

Note

If `opts$READ.TXT=T` and `opts$READ.NROW>0`, then only the first `READ.NROW` records are read from `opts$filename` and from `opts$filetest`. Only the records read are saved to the .Rdata files. This means: when starting `tdmReadData` again with `opts$READ.TXT=F`, not more than these `READ.NROW` records can be retrieved from the relevant .Rdata files. To have again the full data sets on the .Rdata files, you need to invoke `tdmReadData` at least once with `opts$READ.TXT=T` and `opts$READ.NROW=-1`.

If, on the other hand, the full data sets are stored in the .Rdata files, then `opts$READ.TXT=F` and `opts$READ.NROW=<any value>` allows to select a smaller subset of records from the .Rdata files.

Core regression function of TDMR.

Description

tdmRegress is called by `tdmRegressLoop` and returns an object of class `tdmRegre`. It trains a model on training set `d_train` and evaluates it on test set `d_test`. If this function is used for tuning, the test set `d_test` plays the role of a validation set.

Usage

```
tdmRegress(d_train, d_test, d_preproc, response.variables, input.variables,
           opts)
```

Arguments

<code>d_train</code>	training set
<code>d_test</code>	test set, same columns as training set
<code>d_preproc</code>	data used for preprocessing. May be NULL, if no preprocessing is done (<code>opts\$PRE.SFA=="none"</code> and <code>opts\$PRE.PCA=="none"</code>). If preprocessing is done, then <code>d_preproc</code> is usually all non-validation data.
<code>response.variables</code>	name of column which carries the target variable - or - vector of names specifying multiple target columns (these columns are not used during prediction, only for evaluation)
<code>input.variables</code>	vector with names of input columns
<code>opts</code>	additional parameters [defaults in brackets] SRF.* several parameters for sorted_rf_importance (see <code>tdmModelingUtils.r</code>) RF.* several parameters for RF (Random Forest, defaults are set, if omitted) SVM.* several parameters for SVM (Support Vector Machines, defaults are set, if omitted) filename data.title MOD.method ["RF"] the main training method ["RF" "SVM" "LM"]: use [Random forest SVM linear model] for the main model MOD.SEED =NULL: set the RNG to system time as seed (different RF trainings) =any value: set the random number seed to this value (+i) to get reproducible random numbers. In this way, the model training part (RF, NNET, ...) gets always a fixed seed. (see also TST.SEED in <code>tdmRegressLoop</code>) OUTTRAFO [NULL] string, apply a transformation to the output variable

fct.postproc [NULL] name of a user-def'd function for postprocessing of predicted output
 gr.log =FALSE (def): make scatter plot as-is, =TRUE: transform output x with $\log(x+1)$ (x should be nonnegative)
 GD.DEVICE if !="non", then make a pairs-plot of the 5 most important variables and make a true-false bar plot
 VERBOSE [2] =2: most printed output, =1: less, =0: no output

Value

res, an object of class tdmRegre, this is a list containing

d_train	training set + predicted class column(s)
d_test	test set + predicted target output
allRMAE	data frame with columns = (rmae.train, rmae.test, theil.train, theil.test, ...) and rows = response variables. Here Theil's U is based on RMAE (relative mean absolute error).
allRMSE	data frame with columns = (rmse.train, rmse.test, theil.train, theil.test, ...) and rows = response variables. Here Theil's U is based on RMSE (root mean square error).
lastModel	the last model built (e.g. the last Random Forest in the case of MOD.method=="RF")
opts	parameter list from input, some default values might have been added

The item lastModel is specific for the *last* model (the one built for the last response variable in the last run and last fold)

Author(s)

Wolfgang Konen, FHK, Sep'2009 - Jun'2012

See Also

[print.tdmRegre](#) [tdmRegressLoop](#) [tdmClassifyLoop](#)

Examples

```
### This example shows a simple data mining process (phase 1 of TDMR) for regression on
### dataset iris.
### The data mining process in tdmRegress calls randomForest as the prediction model.
### It is called for 2 response variables. Therefore, the data frames allRMAE and allRMSE
### have 2 rows.
###
opts=tdmOptsDefaultsSet()           # set all defaults for data mining process
gdObj <- tdmGraAndLogInitialize(opts); # init graphics and log file

data(iris)
response.variables=c("Petal.Length", "Petal.Width")           # names, not data (!)
input.variables=setdiff(names(iris), response.variables)
opts$rgain.type="rmae"
```

```

opts$NRUN=1

idx_train = sample(nrow(iris))[1:110]
d_train=iris[idx_train,]
d_vali=iris[-idx_train,]
res <- tdmRegress(d_train,d_vali,NULL,response.variables,input.variables,opts)

print(res$allRMAE)
print(res$allRMSE)

```

tdmRegressLoop	<i>Core regression double loop of TDMR returning a TDMregressor object.</i>
----------------	---

Description

tdmRegressLoop contains a double loop (opts\$NRUN and CV-folds) and calls [tdmRegress](#). It is called by all R-functions main_*. It returns an object of class [TDMregressor](#).

Usage

```
tdmRegressLoop(dset, response.variables, input.variables, opts)
```

Arguments

dset	the data frame for which cvi is needed
response.variables	name of column which carries the target variable - or - vector of names specifying multiple target columns (these columns are not used during prediction, only for training and for evaluating the predicted result)
input.variables	vector with names of input columns
opts	a list from which we need here the following entries <ul style="list-style-type: none"> NRUN number of runs (outer loop) TST.SEED =NULL: leave the random number seed as it is. =any value: set the random number seed to this value to get reproducible random numbers and thus reproducible training-test-set-selection. (only relevant in case TST.kind=="cv" or "rand") (see also MOD.SEED in tdmClassify) TST.kind how to create cvi, handed over to tdmModCreateCVindex. If TST.kind="col", then cvi is taken from dset[,opts\$TST.col]. GD.RESTART [TRUE] =TRUE/FALSE: do/don't restart graphic devices GRAPHDEV ["non" other]

Value

result, an object of class `TDMregressor`, this is a list with results, containing

<code>opts</code>	the <code>res\$opts</code> from <code>tdmRegress</code>
<code>lastRes</code>	last run, last fold: result from <code>tdmRegress</code>
<code>R_train</code>	RMAE / RMSE on training set (vector of length <code>NRUN</code>), depending on <code>opts\$rgain.type=="rmae"</code> or <code>"rmse"</code>
<code>S_train</code>	RMSE on training set (vector of length <code>NRUN</code>)
<code>T_train</code>	Theil's U for RMAE on training set (vector of length <code>NRUN</code>)
<code>*_test</code>	— similar, with test set instead of training set —
<code>Err</code>	a data frame with as many rows as <code>opts\$NRUN</code> and columns = (<code>rmae.trn</code> , <code>rmse.trn</code> , <code>made.trn</code> , <code>rmae.theil.trn</code> , <code>ntrn</code> , <code>rmae.tst</code> , <code>rmse.tst</code> , <code>made.tst</code> , <code>rmae.theil.tst</code> , <code>ntst</code>)
<code>predictions</code>	last run: data frame with dimensions [<code>nrow(dset)</code> , <code>length(response.variable)</code>]. In case of CV, all validation set predictions (for each record in <code>dset</code>), in other cases mixed validation / train set predictions.

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), FHK, Sep'2010 - Jun'2012

See Also

[tdmRegress](#), [tdmClassifyLoop](#), [tdmClassify](#)

Examples

```
### ----- demo/demo00-1regress.r -----
### This demo shows a simple data mining process (phase 1 of TDMR) for regression on
### dataset iris.
### The data mining process in tdmRegressLoop calls randomForest as the prediction model.
### It is called opts$NRUN=2 times with different random train-validation set splits.
### Therefore data frame result$Err has 2 rows.
###
opts=tdmOptsDefaultsSet()           # set all defaults for data mining process
gdObj <- tdmGraAndLogInitialize(opts); # init graphics and log file

data(iris)
response.variables="Petal.Length"    # names, not data (!)
input.variables=setdiff(names(iris),"Petal.Length")
opts$rgain.type="rmae"

result = tdmRegressLoop(iris,response.variables,input.variables,opts)

print(result$Err)
```

tdmRegressSummary	<i>Print summary output for result from tdmRegressLoop and add result\$y.</i>
-------------------	---

Description

result\$y is "OOB RMAE" on training set for methods RF or MC.RF. result\$y is "RMAE" on test set (=validation set) for all other methods. result\$y is the quantity which the tuner seeks to minimize.

Usage

```
tdmRegressSummary(result, opts, dset = NULL)
```

Arguments

result	return value from a prior call to tdmRegressLoop , an object of class TDMregressor.
opts	a list from which we need here the following entries NRUN number of runs (outer loop) method VERBOSE dset [NULL] if !=NULL, attach it to result
dset	[NULL] if not NULL, add this data frame to the return value (may cost a lot of memory!)

Value

result, an object of class TDMregressor, with result\$y, result\$sd.y (and optionally also result\$dset) added

Author(s)

Wolfgang Konen, FHK, Sep'2010 - Oct'2011

See Also

[tdmRegress](#), [tdmRegressLoop](#), [tdmClassifySummary](#)

tdmROCR.TDMclassifier *Interactive plot of ROC, lift or other charts for a TDMclassifier object.*

Description

Brings up a [twiddle](#) user interface, where the user may select a part of the dataset ("training" or "validation"), a run number (if `Opts(x)$NRUN>1`) and a type-of-chart, see [tdmROCRbase](#) for details. Using [tdmROCRbase](#), the appropriate chart is plotted on the current graphics device.

Usage

```
## S3 method for class 'TDMclassifier'
tdmROCR(x, ...)
```

Arguments

`x` return value from a prior call to [tdmClassifyLoop](#), an object of class `TDMclassifier`.
`...` – currently not used –

Value

The area under the curve plotted most recently.

Note

Side effect: For each chart, calculate and print the area between the curve and the bottom line ($y=1.0$ for `typ=="lift"`, $y=0.0$ else).

See Also

[tdmClassifyLoop](#) [tdmROCRbase](#)

Examples

```
## Not run:
  setwd(paste(find.package("TDMR"), "demo02sonar", sep="/"));
  source("main_sonar.r");
  result = main_sonar();
  tdmROCR(result);

## End(Not run)
```

tdmROCRbase

*Single plot of ROC, lift or other chart for a [TDMclassifier](#) object.***Description**

Single plot of ROC, lift or other chart for a [TDMclassifier](#) object.

Usage

```
tdmROCRbase(x, dataset = "validation", nRun = 1, typ = "ROC",
            noPlot = FALSE, ...)
```

Arguments

x	return value from a prior call to tdmClassifyLoop , an object of class TDMclassifier .
dataset	["validation"] which part of the data to use, either "training" or "validation"
nRun	[1] if x contains multiple runs, which run to show (1,...,Opts(x)\$NRUN)
typ	["ROC"] which chart type, one out of ("ROC","lift","precRec") for (ROC, lift, precision-recall)-chart (see performance in package ROCR for more details): <ul style="list-style-type: none"> • "ROC": receiver operating curve, TPR vs. FPR, with $TPR=TP/(TP+FN)=TP/P$ and $FPR=FP/(FP+TN)=FP/N$ (true and false positive rate). • "lift": lift chart, LIFT vs. RPP, with $LIFT=TPR/RPR$ with random positive rate $RPR=P/(P+N)$ and $RPP=(TP+FP)/(P+N)$ (rate of pos. predictions). • "precRec": precision-recall-chart, PREC vs. RECALL, with $PREC=TP/(TP+FP)$ and $RECALL=TP/P$ (same as TPR).
noPlot	[FALSE] if TRUE, suppress the plot, return only the area under curve
...	currently not used

Value

The area between the curve and the bottom line $y=0.0$ in the case of `typ=="ROC" | typ=="precRec"` or the area between the curve and the bottom line $y=1.0$ in the case of `typ=="lift"`.

If object x does not contain a prediction score, a warning is issued and the return value is NULL.

See Also

[tdmClassifyLoop](#) [tdmROCR](#).[TDMclassifier](#)

Examples

```
##*# ----- demo/demo06ROCR.r -----
##*# Run task SONAR with "area under ROC curve" as performance measure (rgain.type="arROC").
##*# Other settings are similar to demo01-1sonar.r (level 1 of TDMR).
##*# Finally, plot ROC curve for validataion data set and
##*#           plot lift chart for training data set
##*#
```

```

path <- paste(find.package("TDMR"), "demo02sonar", sep="/");
#path <- paste("../inst", "demo02sonar", sep="/");

## set the elements of list opts. See ?tdmOptsDefaultsSet
## for a complete list of all default settings and explanatory comments
opts = tdmOptsDefaultsSet();
opts$filename = "sonar.txt"
opts$READ.CMD = "readCmdSonar(filename,opts)" # def'd in main_sonar.r
opts$data.title <- "Sonar Data";
opts$rgain.type <- "arROC";

source(paste(path,"start_sonar.r",sep="/"),chdir=TRUE); # contains: result=main_sonar(opts);

tdmGraphicNewWin(opts);
cat("Area under ROC-curve for validation data set: ",
    tdmROCRbase(result),"\n"); # side effect: plot ROC-curve
tdmGraphicNewWin(opts);
cat("Area under lift curve for training data set: ",
    tdmROCRbase(result,dataset="training",typ="lift"),"\n"); # side effect: plot lift chart
#setwd(oldwd);

```

tdmSplitTestData	<i>Read and split the task data and return a TDMdata object.</i>
------------------	--

Description

Read the task data using [tdmReadData](#) and split them into a test part and a training/validation-part.

Usage

```
tdmSplitTestData(opts, tdm, nExp = 0, dset = NULL)
```

Arguments

opts	<p>a list from which we need here the elements</p> <ul style="list-style-type: none"> • READ.INI: [T]=T: do read and split, =F: return NULL • READ.*: other settings for tdmReadData • filename: needed for tdmReadData • filetest: needed for tdmReadData • TST.testFrac: [0.1] set this fraction of the daa aside for testing • TST.COL: string with name for the partitioning column, if tdm\$umode is not "SP_T". (If tdm\$umode=="SP_T", then TST.COL="tdmSplit" is used.)
tdm	<p>a list from which we need here the elements</p> <ul style="list-style-type: none"> • mainFile: if not NULL, set working dir to dir(mainFile) before executing tdmReadData • umode: ["RSUB" "CV" "TST" "SP_T"], how to divide in training/validation data for tuning and test data for the unbiased runs

- `SPLIT.SEED`: if not NULL, set random number generator to this value → deterministic test set split
- `stratified`: [NULL] if not NULL, do the split by stratified sampling (at least one record of each class level found in `dset[, tdm$stratified]` shall appear in the train-vali-set, recommended for classification)

<code>nExp</code>	[0] experiment counter, used to select a reproducible different seed, if <code>tdm\$SPLIT.SEED!=NULL</code>
<code>dset</code>	[NULL] if non-NULL, reading of <code>dset</code> is skipped and the given data frame <code>dset</code> is used.

Value

`dataObj`, either NULL (if `opts$READ.INI==FALSE`) or an object of class `TDMdata` containing

<code>dset</code>	a data frame with the complete data set
<code>TST.COL</code>	string, the name of the column in <code>dset</code> which has a 1 for records belonging to the test set and a 0 for train/vali records. If <code>tdm\$umode=="SP_T"</code> , then <code>TST.COL="tdmSplit"</code> , else <code>TST.COL=opts\$TST.COL</code> .
<code>filename</code>	<code>opts\$filename</code> , from where the data were read

Use the accessor functions `dsetTrnVa.TDMdata` and `dsetTest.TDMdata` to extract the train/vali and the test data, resp., from `dataObj`.

Known caller: `tdmBigLoop`

Author(s)

Wolfgang Konen (<wolfgang.konen@fh-koeln.de>), 2012 - 2013

See Also

`dsetTrnVa.TDMdata`, `dsetTest.TDMdata`, `tdmReadData`, `tdmBigLoop`

<code>tdmStartSpot</code>	<i>Function called by <code>spot</code> to evaluate a DM task during a SPOT tuning run.</i>
---------------------------	---

Description

This function is called by `spot` for evaluations during a tuning run. It accumulates in `spotConfig$alg.currentResult` the RES data frame of all evaluations and in `spotConfig$alg.currentBest` the BST data frame of the so far best solution.

Usage

```
tdmStartSpot(spotConfig)
```

Arguments

- spotConfig the list of configurations for [SPOT](#). Besides the usual [SPOT](#) settings, this list has to contain an element `tdm` with the mandatory elements
- `tdm$mainFunc`: the function name of the DM task to execute (e.g. "main_sonar")
It is expected that `mainFunc` returns a list `result` with the element `result$y`, the quantity to be minimized by [SPOT](#) or other tuners.
- Optionally, `tdm` may define
- `tdm$mainFile`: e.g. "myDir/main_sonar.r", the R file of the DM task to source and `dirname(tdm$mainFile)` is the directory where `tdm$mainFunc` is executed.

Details

[spot](#) is invoked from [spotTuner](#) or [lhdTuner](#). The latter is simply a [SPOT](#)-run with all budget devoted to the initial design.

Value

spotConfig

Author(s)

Wolfgang Konen

See Also

[spot](#), [spotTuner](#), [lhdTuner](#), [tdmDispatchTuner](#)

unbiasedRun

Perform unbiased runs with best-solution parameters.

Description

Read the best solution of a parameter-tuning run from `envT$bst`, execute with these best parameters the function `tdm$mainFunc` (usually a classification or regression machine learning task), to see whether the result quality is reproducible on independent test data or on independently trained models.

Usage

```
unbiasedRun(confFile, envT, dataObj = NULL, finals = NULL, umode = "RSUB",
  withParams = FALSE, tdm = NULL)
```

Arguments

confFile	the .conf filename, e.g. "appAcid_02.conf"
envT	environment, from which we need the objects
	bst data frame containing best results (merged over repeats)
	res data frame containing all results
	theTuner ["spot"] string
	spotConfig [NULL] a list with SPOT settings. If NULL, try to read spotConfig from confFile.
dataObj	[NULL] contains the pre-fetched data from which we use here the test-set part. If NULL, set it to <code>tdmSplitTestData(opts, tdm)</code>
finals	[NULL] a one-row data frame to which new columns with final results are added
umode	["RSUB" (default) "CV" "TST" "SP_T"], how to divide in training and test data for the unbiased runs: "RSUB" random subsampling into $(1-\text{tdm}\$TST.\text{testFrac})\%$ training and $\text{tdm}\$TST.\text{testFrac}\%$ test data "CV" cross validation (CV) with <code>tdm\$run</code> folds "TST" all data in <code>opts\$filename</code> are used for training, all data in <code>opts\$filetest</code> for testing "SP_T" 'split_test': prior to tuning, the data set was split by random subsampling into $\text{tdm}\$TST.\text{testFrac}\%$ test and $(1-\text{tdm}\$TST.\text{testFrac})\%$ training-vali data, tagged via column "tdmSplit". Tuning was done on training-vali data. Now we use column "tdmSplit" to select the test data for unbiased evaluation. Training during unbiased evaluation is done on a fraction <code>tdm\$TST.trnFrac</code> of the training-vali data
withParams	[FALSE] if =TRUE, add columns with best parameters to data frame finals (should be FALSE, if different runs have different parameters)
tdm	a list with TDM settings from which we use here the elements mainFunc the function to be called for unbiased evaluations mainFile change to the directory of mainFile before starting mainFunc nrun [5] how often to call the unbiased evaluation ifold [10] how many folds in CV (only relevant for umode="CV") TST.testFrac [0.2] test set fraction (only relevant for umode="RSUB" or ="SP_T") The defaults in '[...]' are set by <code>tdmDefaultsFill</code> , if they are not defined on input.

Value

finals a one-row data frame with final results

Note

Side Effects: The list result, an object of class `TDMclassifier` or `TDMregressor` as returned from `tdm$mainFunc` is written onto `envT$result`.

If `envT$spotConfig` is NULL, it is constructed from `confFile`.

If `spotConfig$opts` (list with all parameter settings for the DM task) is `NULL`, we try to read it from `spotConfig$io.apdFileName`. This will issue a warning '... might not work in parallel mode', but is perfectly fine for non-parallel mode.

Author(s)

Wolfgang Konen, FHK, 2010 - 2013

See Also

[tdmBigLoop](#), [TDMclassifier](#), [TDMregressor](#)

Examples

```
## Load the best results obtained in a prior tuning for the configuration
## "sonar_04.conf" with tuning method "spot". The result envT from a prior
## run of tdmBigLoop with this .conf is read from demo02sonar/demoSonar.RData.
## Run task main_sonar again with these best parameters, using the default
## settings from tdmDefaultsFill: umode="RSUB", tdm$nrnrun=5 and tdm$TST.testFrac=0.2.
oldwd <- getwd();
## The best results are read from demo02sonar/demoSonar.RData relative to
## the TDMR package directory.
setwd(paste(find.package("TDMR"), "demo02sonar", sep="/"));
load("demoSonar.RData");
source("main_sonar.r");
finals <- unbiasedRun("sonar_04.conf", envT, tdm=envT$tdm);
print(finals);
setwd(oldwd);
```

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