

# Package

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**Type** Package

**Title** Machine Learning Automation

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## Description

SmartML is a meta learning-based framework for automated selection and hyperparameter tuning for machine learning algorithms. Being meta-learning based, the framework is able to simulate the role of the machine learning expert. In particular, the framework is equipped with a continuously updated knowledge base that stores information about the meta-features of all processed datasets along with the associated performance of the different classifiers and their tuned parameters. Thus, for any new dataset, SmartML automatically extracts its meta features and searches its knowledge base for the best performing algorithm to start its optimization process. In addition, SmartML makes use of the new runs to continuously enrich its knowledge base to improve its performance and robustness for future runs.

**License** GPL-2

**Encoding** UTF-8

**LazyData** true

## Imports

devtools,R.utils,stats,httr,mice,RCurl,tictoc,e1071,caret,mlbench,fastICA,RMySQL,BBmisc,rjson,ggplot2,RWeka,far

**Suggests** knitr

**RoxygenNote** 6.1.1

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autoRLearn

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Run smartML function for automatic Supervised Machine Learning.

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## Description

Run the smartML main function for automatic classifier algorithm selection, and hyper-parameter tuning.

## Usage

```
autoRLearn(maxTime, directory, testDirectory, classCol = "class",
  selectedFeats = c(), vRatio = 0.1, preProcessF = "N",
  featuresToPreProcess = c(), nComp = NA, nModels = 3, option = 2,
  featureTypes = c(), interp = 0, missingVal = c("NA", "?", " "),
  missingOpr = 0)
```

## Arguments

maxTime	Float of the maximum time budget for reading dataset, preprocessing, calculating meta-features, Algorithm Selection & hyper-parameter tuning process only in minutes(Excluding Model Interpretability) - This is applicable in case of Option = 2 only.
directory	String of the training dataset directory (SmartML accepts file formats arff/(csv with columns headers) ).
testDirectory	String of the testing dataset directory (SmartML accepts file formats arff/(csv with columns headers) ).
classCol	String of the name of the class label column in the dataset (default = 'class').
selectedFeats	Vector of numbers of features columns to include from the training set and ignore the rest of columns - In case of empty vector, this means to include all features in the dataset file (default = c()).
vRatio	Float of the validation set ratio that should be splitted out of the training set for the evaluation process (default = 0.1 -> 10%).
preProcessF	string containing the name of the preprocessing algorithm (default = 'N' -> no preprocessing): <ul style="list-style-type: none"> <li>• "boxcox" - apply a Box-Cox transform and values must be non-zero and positive in all features,</li> <li>• "yeo-Johnson" - apply a Yeo-Johnson transform, like a BoxCox, but values can be negative,</li> <li>• "zv" - remove attributes with a zero variance (all the same value),</li> <li>• "center" - subtract mean from values,</li> <li>• "scale" - divide values by standard deviation,</li> <li>• "standardize" - perform both centering and scaling,</li> <li>• "normalize" - normalize values,</li> <li>• "pca" - transform data to the principal components,</li> <li>• "ica" - transform data to the independent components.</li> </ul>

featuresToPreProcess	Vector of number of features to perform the feature preprocessing on - In case of empty vector, this means to include all features in the dataset file (default = c()) - This vector should be a subset of selectedFeats.
nComp	Integer of Number of components needed if either "pca" or "ica" feature preprocessors are needed.
nModels	Integer representing the number of classifier algorithms that you want to select based on Meta-Learning and start to tune using Bayesian Optimization (default = 3).
option	Integer representing either Classifier Algorithm Selection is needed only = 1 or Algorithm selection with its parameter tuning is required = 2 which is the default value.
featureTypes	Vector of either 'numerical' or 'categorical' representing the types of features in the dataset (default = c() -> any factor or character features will be considered as categorical otherwise numerical).
interp	Boolean representing if model interpretability (Feature Importance and Interaction) is needed or not (default = 0) This option will take more time budget if set to 1.
missingVal	Vector of strings representing the missing values in dataset (default: c('NA', '?', ' ')).
missingOpr	Boolean variable represents either delete instances with missing values or apply imputation using "MICE" library which helps you imputing missing values with plausible data values that are drawn from a distribution specifically designed for each missing datapoint- (default = 0 -> delete instances).

## Value

### List of Results

- "option=1" - Chosen Classifier Algorithms Names clfs with their parameters configurations params in case of option=2,
- "option=2" - Best classifier algorithm name found clfs with its parameters configuration params, model variable model, performance on TestingSet perf, and Feature Importance interpret\$featImp/Interaction interpret\$Interact plots in case of interpretability interp is needed.

## Examples

```
## Not run:
autoRLearn(1, 'sampleDatasets/car/train.arff', \
'sampleDatasets/car/test.arff', option = 2, preProcessF = 'normalize')

result <- autoRLearn(10, 'sampleDatasets/shuttle/train.arff', 'sampleDatasets/shuttle/test.arff')

## End(Not run)
```

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datasetReader	<i>Read Dataset File into Memory.</i>
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## Description

Read the file of the training and testing dataset, and perform preprocessing and data cleaning if necessary.

## Usage

```
datasetReader(directory, testDirectory, selectedFeats = c(),
  classCol = "class", preProcessF = "N", featuresToPreProcess = c(),
  nComp = NA, missingVal = c("NA", "?", " "), missingOpr = 0)
```

## Arguments

directory	String of the directory to the file containing the training dataset.
testDirectory	String of the directory to the file containing the testing dataset.
selectedFeats	Vector of numbers of features columns to include from the training set and ignore the rest of columns - In case of empty vector, this means to include all features in the dataset file (default = c()).
classCol	String of the name of the class label column in the dataset (default = 'class').
preProcessF	string containing the name of the preprocessing algorithm (default = 'N' -> no preprocessing): <ul style="list-style-type: none"> <li>• "boxcox" - apply a Box-Cox transform and values must be non-zero and positive in all features,</li> <li>• "yeo-Johnson" - apply a Yeo-Johnson transform, like a BoxCox, but values can be negative,</li> <li>• "zv" - remove attributes with a zero variance (all the same value),</li> <li>• "center" - subtract mean from values,</li> <li>• "scale" - divide values by standard deviation,</li> <li>• "standardize" - perform both centering and scaling,</li> <li>• "normalize" - normalize values,</li> <li>• "pca" - transform data to the principal components,</li> <li>• "ica" - transform data to the independent components.</li> </ul>
featuresToPreProcess	Vector of number of features to perform the feature preprocessing on - In case of empty vector, this means to include all features in the dataset file (default = c()) - This vector should be a subset of selectedFeats.
nComp	Integer of Number of components needed if either "pca" or "ica" feature preprocessors are needed.
missingVal	Vector of strings representing the missing values in dataset (default: c('NA', '?', ' ')).
missingOpr	Boolean variable represents either delete instances with missing values or apply imputation using "MICE" library which helps you imputing missing values with plausible data values that are drawn from a distribution specifically designed for each missing datapoint- (default = 0 -> delete instances).

**Value**

List of the TrainingSet Train and TestingSet Test.

**Examples**

```
## Not run:
dataset <- datasetReader('/Datasets/irisTrain.csv', '/Datasets/irisTest.csv')

## End(Not run)
```

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runClassifier

*Fit a classifier model.*


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**Description**

Run the classifier on a training set and measure performance on a validation set.

**Usage**

```
runClassifier(trainingSet, validationSet, params, classifierAlgorithm,
  interp = 0)
```

**Arguments**

trainingSet	Dataframe of the training set.
validationSet	Dataframe of the validation Set.
params	A string of parameter configuration values for the current classifier to be tuned (parameters are separated by #) and can be obtained from params out of resulted list after running autoRLearn function.
classifierAlgorithm	String of the name of classifier algorithm used now. <ul style="list-style-type: none"> <li>• "svm" - Support Vector Machines from e1071 package,</li> <li>• "naiveBayes" - naiveBayes from e1071 package,</li> <li>• "randomForest" - randomForest from randomForest package,</li> <li>• "lmt" - LMT Weka classifier trees from RWeka package,</li> <li>• "lda" - Linear Discriminant Analysis from MASS package,</li> <li>• "j48" - J48 Weka classifier Trees from RWeka package,</li> <li>• "bagging" - Bagging Classifier from ipred package,</li> <li>• "knn" - K nearest Neighbors from FNN package,</li> <li>• "nnet" - Simple neural net from nnet package,</li> <li>• "fda" - Flexible discriminant Analysis from MDA package,</li> <li>• "C50" - C50 decision tree from C5.0 package,</li> <li>• "rpart" - rpart decision tree from rpart package,</li> <li>• "rda" - regularized discriminant analysis from klaR package,</li> <li>• "plsda" - Partial Least Squares And Sparse Partial Least Squares Discriminant Analysis from caret package,</li> <li>• "glm" - Fitting Generalized Linear Models from stats package,</li> <li>• "deepboost" - deep boost classifier from deepboost package.</li> </ul>
interp	Boolean representing if interpretability is required or not (Default = 0).

**Value**

List of performance on validationSet named perf, model fitted on trainingSet named trainingSet, and interpretability plots named interact in case of interp = 1

**Examples**

```
## Not run:
result1 <- autoRLearn(10, 'sampleDatasets/shuttle/train.arff', 'sampleDatasets/shuttle/test.arff')
dataset <- datasetReader('/Datasets/irisTrain.csv', '/Datasets/irisTest.csv')
result2 <- runClassifier(dataset$Train, dataset$Test, result1$params, result1$clfs)

## End(Not run)
```

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