Package 'metaEnsembleR'

October 13, 2022

| Type | Package |
|--------|---|
| Title | Automated Intuitive Package for Meta-Ensemble Learning |
| Versio | on 0.1.0 |
| Date | 2020-10-27 |
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| | iption Extends the base classes and methods of 'caret' package for integration of base learners. The user can input the number of different base learners, and specify the final learner, along with the train-validation-test data partition split ratio. The predictions on the unseen new data is the resultant of the ensemble metalearning https://machinelearningmastery.com/stacking-ensemble-machine-learning-with-python/ of the heterogeneous learners aimed to reduce the generalization error in the predictive models. It significantly lowers the barrier for the practitioners to apply heterogeneous ensemble learning techniques in an amateur fashion to their everyday predictive problems. |
| | se GPL (>= 2) |
| Encod | ling UTF-8 |
| Needs | Compilation no |
| Repos | itory CRAN |
| Impor | ts caret, ggplot2, graphics, e1071, gbm, randomForest |
| Depen | nds gridExtra |
| Sugge | sts knitr, R.rsp |
| Vigne | tteBuilder R.rsp |
| Roxyg | genNote 7.1.1 |
| Date/I | Publication 2020-11-19 09:10:05 UTC |
| R to | pics documented: |
| | ensembler.classifier |
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ensembler.classifier Ensemble Classifiers Training & Prediction, Model Result Evaluation

Description

This function uses the base learners, and the final top layer learner to produce an ensemble prediction. The user can input the number of base learners and specify the final learner along with the train-validation-test data partition split ratio. The predictions on the unseen new data is the resultant of the ensemble meta-learning of the heterogeneous learners aimed to reduce the generalization error in the predictive models. Functions from **caret** are used for training and prediction of the base learners and the final learner.

Usage

```
ensembler.classifier(data
  , outcomeVARIABLEINDEX
  , IndividualModels
  , TopLayerModel
  , dstr
  , dsv
  , dst
  , unseen_new_data
)
```

Arguments

data Data to be used for training, validation, and test.

outcomeVARIABLEINDEX

Index of the response/outcome variable.

IndividualModels

Training of base learners.

TopLayerModel Top layer final learner.

dstr Training data split ratio.

dsv Validation data split ratio.

dst Testing data split ratio.

unseen_new_data

Prediction on unseen new data.

Value

data Complete data or the indexed data.

outcomeVARIABLEINDEX

Integer value of the response/outcome variable index.

IndividualModels

A vector of base learners or standalone individual learner.

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| TopLayerModel | Name of the Top layer learner. | |
|-----------------|--|--|
| dstr | fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal to 1. | |
| dsv | fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal to $1.$ | |
| dst | fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal to 1 . | |
| unseen_new_data | | |

Unseen new data provided either as a csv file or imported dataframe.

Note

The function ensembler_classifier), returns a list with the following elements: ensembler_return[1]Test data predictions. ensembler_return[2]Prediction labels. ensembler_return[3]Model Result. ensembler_return[4]Unseen data predictions.

Author(s)

Ajay Arunachalam

Examples

```
library("metaEnsembleR")
attach(iris)
data("iris")
unseen_new_data_testing <- iris[130:150,]</pre>
#write.csv(unseen_new_data_testing
        , 'unseen_check.csv'
               , fileEncoding = 'UTF-8'
#
#
                       , row.names = FALSE)
ensembler_return <- ensembler.classifier(iris[1:130,]</pre>
     , c('rpart') #c('treebag','rpart')
        , 'rf'
                            # 'gbm'
            , 0.60
                , 0.20
                   , 0.20
                         , unseen_new_data_testing)
#ensembler_return <- ensembler.classifier(iris[1:130,]</pre>
#
#
      , c('treebag','rpart')
           , 'gbm'
#
#
                , 0.60
                 , 0.20
#
                    , 0.20
#
                         , read.csv('./unseen_check.csv'))
testpreddata <- data.frame(ensembler_return[1])</pre>
table(testpreddata$actual_label)
```

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```
#### Performance comparison #####
modelresult <- ensembler_return[3]
modelresult
act_mybar <- qplot(testpreddata$actual_label, geom="bar")
act_mybar
pred_mybar <- qplot(testpreddata$predictions, geom='bar')
pred_mybar
act_tbl <- tableGrob(t(summary(testpreddata$actual_label)))
pred_tbl <- tableGrob(t(summary(testpreddata$predictions)))
#ggsave("testdata_actual_vs_predicted_chart.pdf",grid.arrange(act_tbl, pred_tbl))
#ggsave("testdata_actual_vs_predicted_plot.pdf",grid.arrange(act_mybar, pred_mybar))
#### unseen data ###
unseenpreddata <- data.frame(ensembler_return[4])
table(unseenpreddata$unseenpreddata)</pre>
```

ensembler.regression Ensemble Regressor Training & Prediction, Model Result Evaluation

Description

This function uses the base learners, and the final top layer learner to produce an ensemble prediction. The user can input the number of base learners and specify the final learner along with the train-validation-test data partition split ratio. The predictions on the unseen new data is the resultant of the ensemble meta-learning of the heterogeneous learners aimed to reduce the generalization error in the predictive models. Functions from **caret** are used for training and prediction of the base learners and the final learner.

Usage

```
ensembler.regression(data
  , outcomeVARIABLEINDEX
  , IndividualModels
  , TopLayerModel
  , dstr
  , dsv
  , dst
  , unseen_new_data
)
```

Arguments

```
data Data to be used for training, validation, and test. outcomeVARIABLEINDEX
```

Index of the response/outcome variable.

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IndividualModels

Training of base learners.

TopLayerModel Top layer final learner.

dstr Training data split ratio.

dsv Validation data split ratio.

dst Testing data split ratio.

unseen_new_data

Prediction on unseen new data.

Value

data Complete data or the indexed data.

outcomeVARIABLEINDEX

Integer value of the response/outcome variable index.

IndividualModels

A vector of base learners or standalone individual learner.

TopLayerModel Name of the Top layer learner.

dstr fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal

to 1.

dsv fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal

to 1.

dst fraction integer. Not greater than 1. Total sum of dstr, dsv, dst should be equal

to 1.

unseen_new_data

Unseen new data provided either as a csv file or imported dataframe.

Note

The function ensembler.regression), returns a list with the following elements: ensembler_return[1]Test data predictions. ensembler_return[2]Prediction values. ensembler_return[3]Model Result. ensembler_return[4]Unseen data predictions.

Author(s)

Ajay Arunachalam

Examples

```
library("metaEnsembleR")
data("rock")
unseen_rock_data <- rock[30:48,]
ensembler_return <- ensembler.regression(rock[1:30,]
    , 4
     ,c('lm')
     , 'rf'
     , 0.40</pre>
```

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```