Homework on Supervised Learning Dolores Romero Morales

The minicreditgerman dataset (creditgerman.txt) contains 20 explanatory variables and 750 observations. The class informs us whether the client was a good payer ('1') or not ('2'):

Using the best practices in Data Science, address the following questions making sure to describe the modeling approach used:

- (i) Use Logistic Regression to model the creditgerman data.
- (ii) Use Support Vector Machines to model the creditgerman data.
- (iii) Use Classification Trees to model the creditgerman data.
- (iv) Use Random Forests to model the creditgerman data.
- (v) Use k-Nearest Neighbors to model the creditgerman data.

(vi) Use the models built in Questions 1(i)-1(v) to predict the class membership of the objects in testcreditgerman.txt.

(vii) ...

(viii) ...

Answers

Note1: With the workshop and homework answers provided in CANVAS, you should be able to code in R the models. Note2: I kindly remind you that you have been asked to describe the modeling approach Note3: I have kicked off the process adding some lines of code here below. You can fill in the remaining lines. Note4: Recall that you need to upload the corresponding packages

#these lines are to read the two files provided
mymini <- read.table(file.choose(),header=TRUE, stringsAsFactors=TRUE)
mytesting <- read.table(file.choose(),header=TRUE, stringsAsFactors=TRUE)</pre>

#these lines are to inspect the nature of the variables
str(mymini)
str(mytesting)

#these lines are to convert the target variable into a factor and the categorical variables into dummies
library(dummies)
myXmini <- mymini[,-21]
myXminiD <- dummy.data.frame(myXmini, sep= ".")
classmembership <- as.factor(mymini\$classmembership)
myMtesting <- mytesting[,-21]
myXtestingD <- dummy.data.frame(myXtesting, sep= ".")
classmembership <- as.factor(mytesting\$classmembership)
mytestingD <- cbind(myXtestingD, classmembership)
mytestingD <- cbind(myXtestingD, classmembership)</pre>

#these lines are to understand in which position is our target variable now #we can see that we know have 61 explanatory variables and our response variable is in column 62 dim(myminiD)

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#these lines are to normalize the continuous and the integer data
myminiNORM<- myminiD
mytestingNORM<- mytestingD
for(i in 1:length(colnames(myminiD))-1) {
    if(class(myminiD[,i]) == "numeric" || class(myminiD[,i]) == "integer") {
        minimum<-min(myminiD[,i])
        maximum<-max(myminiD[,i])
        myminiNORM[,i] <- as.vector(scale(myminiD[,i],center=minimum,scale=maximum-minimum))
        mytestingNORM[,i] <- as.vector(scale(mytestingD[,i],center=minimum,scale=maximum-minimum))
    }
}</pre>
```

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summary(myminiNORM)
summary(mytestingNORM)
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#this line is to run a Logistic Regression model
#please note that the glm function can handle numeric as well as categorical variables, without you needing to transform
them yourself, although internally, the transformation is occurring.
LRmodel <- glm(classmembership~ ., family='binomial', myminiNORM)
#this line is to eliminate irrelevant variables
LRmodelR <- step(LRmodel)</pre>

#this line is to predict on another sample, note that we need to do a bit of more work as glm() is a function that can be
used for regression and for classification
#we find first the probabilities and then we find the predicted class
#remember that the classes are "1" and "2"
probabilitiesLR <- predict(LRmodelR, mytestingNORM[,-62],type= "response")</pre>

predictionLR <- ifelse(probabilitiesLR > 0.5, "2", "1") classificationtable <- table(pred= predictionLR, mytestingNORM[,62]) acctestLR <- sum(diag(classificationtable))/sum(classificationtable) acctestLR

#lines to run SVM models with several kernels

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#lines to run Classification Trees models and to prune them

#lines to run Random Forests models

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#lines to run k-NN models

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