Data Analytics Lab 08/10/19

rm(list=ls())

Step 1 - Import the excel file and checking content of the data

Convert the dataset into csv file, Read.csv tells R to read csv file.

Data <- read.csv("DT-Credit.csv", header=TRUE, sep= ";")

Check distinct categories of Variables using STR function

str(Data)

you need to change some data as factor:

cols <- c(1:2, 4:10, 12:22, 24:32)

Data[cols] <- lapply(Data[cols], factor)

str(Data)

you need to remove the first column:

Data <- Data[,-1]

Check your data:

names(Data)

attach(Data)

Step 2 - Install package rpart, and click on the checkbox in front of rpart library.

Develop the DT model:

DT_Model <-rpart(RESPONSE~., data=Data, control=rpart.control(minsplit=60, minbucket=30, maxdepth=4))

minsplit: the minimum number of observations that must exist in a node for a new split minbucket: the minimum number of observations in any terminal <leaf> node Maxdepth: Maximum depth for any node, with the root node counted as depth 0.

Step 3 - Install packgae partykit:

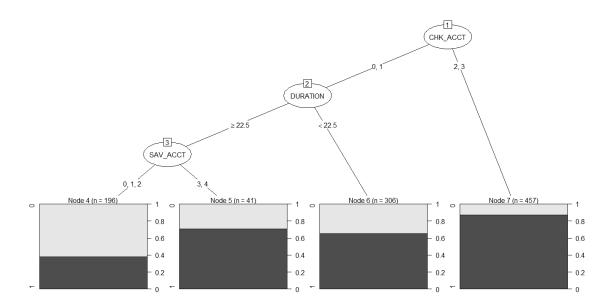
install.packages ("partykit")

library("partykit")

plot(as.party(DT Model))

print(DT Model)

You would get the following output. Describe the results at the end of your scripts.

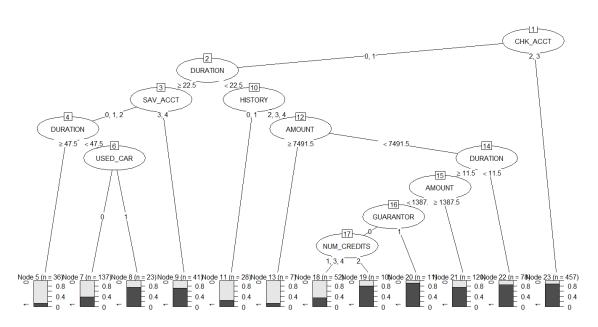


Step 4- Procedure of Pruning

DT_Model2<-rpart(Target~., data=Data)</pre>

Plot(as.party(DT_Model2))

You should get the following output:



The following line fitted tree's CP table (Matrix of Information on optimal pruning given Complexity Parameter). Look where do you see the least error.

print(DT_Model2\$cptable)

The line below automatically picks up the least error tree

opt <- which.min(DT_Model2\$cptable [, "xerror"])

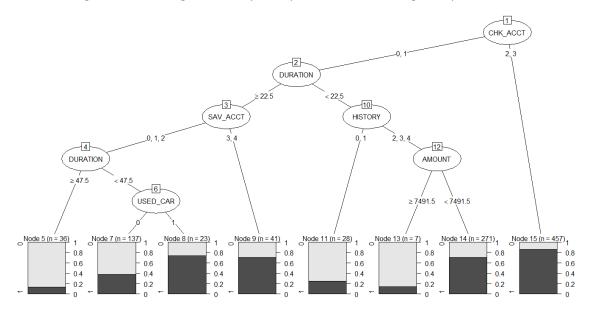
Step 5 - Pruning the tree to the least xerror

cp <- DT Model2\$cptable [opt,"CP"]

DT_Model_Pruned <- prune(DT_Model2, cp=cp)

plot(as.party(DT_Model_pruned))

You should get the following model. Try to explain the result using the cp table above.



Step 6 - Random Forest

Install the package for Random Forest

install.packages ("randomForest")

library(randomForest)

Run the Model

RF <- randomForest(RESPONSE~.,data=Data)

See the result

print(RF)

See importance of each predictor

importance(RF)

Plot the importance

varImpPlot(RF)

See the error vs. number of trees

plot(RF)