Activity Sheet

Learning outcomes

After solving these exercises, you will understand the following:

- 1. Applying the Decision Trees using C5.0 and CART algorithms to solve classification and regression problems respectively
- 2. Understand and interpret the results generated from each algorithm in R
- 3. Comparison of the model performance in terms of Accuracy for Classification
- 4. Comparison of the model performance in terms of Mean square error/ Root Mean square error for regression

Dataset and Target variables

Use the data set "Part1.csv" and "Part2.csv" to solve these exercises.

- Take the variable "RESPONSE" as target variable for <u>classification</u> and apply C5. 0 algorithm to classify whether a customer is good/bad customer
- Take the variable "AMOUNT" as target variable for <u>regression</u> and apply CART algorithm to predict how much a customer is eligible for

Note: We have provided r code wherever it is necessary. This code is for your reference only. Do not copy and paste code in R console. To use this code, please ensure you change the data frame names and the variable names as per your working environment.

Steps to follow for Classification

- 1. Go through 'Data_Descripion.csv' file to understand each variable and identify the data type (Numeric/categorical)
- 2. Import both the 'part1' and 'part2' csv files into R
- 3. Merge all two data sets by "obs" and remove the missing values
- 4. Separate the numerical and categorical data
- 5. Type conversion for each of the attribute- All attributes in data_Cat should be factor and data_Num should be numeric
- 6. Discretizing the numeric data using equal width or equal frequency method library(infotheo)
- 7. Let us construct a new data frame that contains <u>all</u> the variables in appropriate type. We need to create a dataframe with all the categorical and adding discretized numeric variables
- 8. Look at the summary of the data frame to check whether all the variables are categorical or not
- 9. Split the data into Training (60% of the total records) and Testing (40% of the records). Use below R code.



- 10. Let us the start building our first model. You need to install the below library. install.packages("C50")
- 11. Apply C.50 model on the training dataset. Use the below R code.

```
library(C50)
```

```
DT C50=C5.0(response~.,data=TrainData,rules=T)
```

#To get the important attributes

```
C5imp(DT_C50,pct=T)
```

- 12. Understand the summary of the rules generated and the important variables summary(dtC50)
- 13. To validate our algorithm, let us apply on the test data and get the results.

```
a=table(TrainData$response, predict(DT_C50, newdata=TrainData, type="class"))
```

a

```
b=table(TestData$response, predict(DT_C50, newdata=TestData, type="class"))
```

accTrain = sum(diag(a))/sum(a)

accTest = sum(diag(b))/sum(b)

Steps to follow for Regression problem

Our objective is to build a prediction model for the target variable 'Amount'. Let us reuse some of r code we had already written in the above exercise.

- 1. Take the final data you had built earlier and remove the amount variable (because this is in the binned format!).
- 2. Get the 'amount' variable, either from initial data frame or from the numeric data frame you generated in at step-4 in the above activity. Add this variable to the data frame you got in the step-1
- 3. Split the data into Training (60% of the total records) and Testing (40% of the records). Use R code at step-8 in the previous activity.
- 4. You need to install 'rpart' library to build the prediction model.
- 5. Use below r code to build the model

library(rpart)

DT_rpart<-rpart(amount~.,data=TrainData1,method="anova")

plot(DT_rpart,main="Regression tree for Amount",margin=0.15,uniform=T)



text(DT_rpart,use.n=T)

DT_rpart_PredTrain=predict(DT_rpart,newdata=TrainData1, type="vector")

A<-data.frame(TrainData1[28],DT_rpart_PredTrain)

DT_rpart_PredTest=predict(DT_rpart, newdata=TestData1,type="vector")

B<-data.frame(TestData1[28],DT_rpart_PredTest)

library(DMwR)

regr.eval(TrainData1[28],DT_rpart_PredTrain)

