Machine Learning Assignment 2

10-02-2022

```
\#\#\#Activating Packages
library(psych)
library(caret)
## Loading required package: ggplot2
## Attaching package: 'ggplot2'
## The following objects are masked from 'package:psych':
##
##
       %+%, alpha
## Loading required package: lattice
library(FNN)
library(class)
##
## Attaching package: 'class'
## The following objects are masked from 'package:FNN':
       knn, knn.cv
##
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
###importing data
getwd()
## [1] "/Users/bharathreddy/Desktop"
setwd("/Users/bharathreddy/Desktop")
bharat <- read.csv("UniversalBank.csv")</pre>
#Eliminating variables [id & zip code] from the dataset
df=subset(bharat, select=-c(ID, ZIP.Code ))
```

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#creating dummies
dummy Education <- as.data.frame(dummy.code(df$Education))</pre>
names(dummy_Education) <- c("Education_1", "Education_2", "Education_3")</pre>
df_without_education <- subset(df, select=-c(Education))</pre>
UBank_data <- cbind(df_without_education, dummy_Education)</pre>
###Data partition
set.seed(1234)
Train Index
                = createDataPartition(UBank data$Personal.Loan, p= 0.6 , list=FALSE)
Train Data
                = UBank data[Train Index,]
Validation_Data = UBank_data[-Train_Index,]
###Generating test data
Test_Data <- data.frame(Age=40 , Experience=10, Income = 84, Family = 2, CCAvg = 2, Education_1 = 0, Ed
\#\#\# {\operatorname{Data}} Normalization
train.norm.df
                 <- Train_Data
valid.norm.df
                 <- Validation_Data
test.norm.df
                 <- Test_Data
maindata.norm.df <- UBank_data</pre>
head(maindata.norm.df)
     Age Experience Income Family CCAvg Mortgage Personal.Loan Securities.Account
## 1 25
                         49
                                 4
                                     1.6
                  1
## 2 45
                 19
                         34
                                     1.5
                                                 0
                                                                0
                                                                                    1
                                 3
## 3 39
                 15
                                     1.0
                                                 0
                                                                0
                                                                                    0
                         11
                                 1
## 4 35
                  9
                        100
                                     2.7
                                                                0
                                                                                    0
## 5 35
                  8
                         45
                                     1.0
                                                 0
                                                                0
                                                                                    0
                                 4
                 13
                         29
                                 4
                                     0.4
                                               155
## 6 37
## CD.Account Online CreditCard Education_1 Education_2 Education_3
## 1
              0
                     0
                                 0
                                              1
                                                          0
## 2
              0
                      0
                                 0
                                                          0
                                                                       0
                                              1
## 3
              0
                      0
                                 0
                                              1
                                                          0
                                                                       0
## 4
              0
                                 0
                      0
                                              0
                                                          0
                                                                       1
## 5
              0
                      0
                                 1
                                              0
                                                                       1
## 6
              0
                      1
                                 0
                                              0
                                                                       1
# use preProcess() from the caret package to normalize .
norm.values <- preProcess(Train_Data[,-7], method=c("center", "scale"))</pre>
train.norm.df[,-7] <- predict(norm.values, Train_Data[,-7]) #Training Data
valid.norm.df [,-7] <- predict(norm.values, Validation_Data[,-7]) #Validation_Data
test.norm.df <- predict(norm.values, Test_Data)#Test Data</pre>
maindata.norm.df[,-7] <- predict(norm.values,UBank_data[,-7]) #Training + Validation data
head(maindata.norm.df)
             Age Experience
                                  Income
                                              Family
                                                          CCAvg
                                                                   Mortgage
## 1 -1.76517597 -1.66184535 -0.5337999 1.3900576 -0.1858029 -0.5622826
## 2 -0.03138168 -0.09955276 -0.8590893 0.5257731 -0.2428548 -0.5622826
```

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## 3 -0.55151996 -0.44672889 -1.3578664 -1.2027959 -0.5281144 -0.5622826
## 4 -0.89827882 -0.96749308 0.5721841 -1.2027959 0.4417682 -0.5622826
## 5 -0.89827882 -1.05428712 -0.6205437 1.3900576 -0.5281144 -0.5622826
## 6 -0.72489939 -0.62031695 -0.9675191 1.3900576 -0.8704259 0.9676351
    Personal.Loan Securities.Account CD.Account
                                                      Online CreditCard Education 1
## 1
                 Λ
                            2.8635153 -0.2570526 -1.1911682 -0.6364096
                                                                          1.1789719
## 2
                 0
                            2.8635153 -0.2570526 -1.1911682 -0.6364096
                                                                          1.1789719
                           -0.3491047 -0.2570526 -1.1911682 -0.6364096
                                                                          1.1789719
## 3
                 0
## 4
                 Λ
                           -0.3491047 -0.2570526 -1.1911682 -0.6364096 -0.8479139
## 5
                 0
                           -0.3491047 -0.2570526 -1.1911682 1.5707913 -0.8479139
## 6
                 0
                           -0.3491047 -0.2570526 0.8392322 -0.6364096 -0.8479139
##
     Education_2 Education_3
## 1 -0.6477981
                 -0.6327928
## 2 -0.6477981
                 -0.6327928
## 3 -0.6477981
                 -0.6327928
## 4 -0.6477981
                   1.5797694
## 5 -0.6477981
                   1.5797694
## 6 -0.6477981
                   1.5797694
###Perforing k-NN classification, using k = 1
set.seed(1234)
prediction <- knn(train = train.norm.df[,-7], test = valid.norm.df[,-7],</pre>
          cl = train.norm.df[,7], k = 1, prob=TRUE)
actual= valid.norm.df$Personal.Loan
prediction_prob = attr(prediction, "prob")
table(prediction, actual)
             actual
                 0
                      1
## prediction
##
            0 1785
                     61
##
                19 135
            1
mean(prediction==actual)
## [1] 0.96
accuracy.df <- data.frame(k = seq(1, 30, 1), accuracy = rep(0, 30))
for(i in 1:30) {
prediction <- knn(train = train.norm.df[,-7], test = valid.norm.df[-7],</pre>
          cl = train.norm.df[,7], k = i, prob=TRUE)
accuracy.df[i,2] <- mean(prediction==actual)</pre>
accuracy.df
       k accuracy
##
## 1
           0.9600
       1
## 2
       2
           0.9520
           0.9590
## 3
       3
## 4
       4
           0.9540
## 5
       5
           0.9540
## 6
       6
           0.9550
## 7
       7
           0.9515
## 8
           0.9525
```

```
## 9
       9
           0.9495
## 10 10
           0.9490
## 11 11
           0.9485
## 12 12
           0.9485
## 13 13
           0.9465
## 14 14
           0.9465
## 15 15
           0.9460
## 16 16
           0.9440
## 17 17
           0.9435
## 18 18
           0.9445
## 19 19
           0.9430
## 20 20
           0.9430
## 21 21
           0.9430
## 22 22
           0.9420
## 23 23
           0.9420
## 24 24
           0.9420
## 25 25
           0.9420
## 26 26
           0.9405
## 27 27
           0.9425
## 28 28
           0.9425
## 29 29
           0.9400
## 30 30
           0.9390
The value of k we choose is 1 as it is given in the question.
####Validation data results using best k value [i.e. k = 1]
set.seed(1234)
prediction <- knn(train = train.norm.df[,-7], test = valid.norm.df[,-7],</pre>
          cl = train.norm.df[,7], k = 1, prob=TRUE)
actual= valid.norm.df$Personal.Loan
prediction_prob = attr(prediction,"prob")
table(prediction, actual)
##
             actual
## prediction
                       1
            0 1785
                      61
##
            1
                19 135
prediction_test <- knn(train = maindata.norm.df[,-7], test = Test_Data,</pre>
          cl = maindata.norm.df[,7], k = 1, prob=TRUE)
head(prediction_test)
## [1] 1
## Levels: 0 1
k-NN model predicted that the new customer will accept a loan offer.
#Partitioning the data into Traning(50%), Validation(30%), Test(20%)
set.seed(1234)
Test_Index_1 = createDataPartition(UBank_data$Age, p= 0.2 , list=FALSE)
Test_Data_1 = UBank_data [Test_Index_1,]
Rem_DATA = UBank_data[-Test_Index_1,]
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Train_Index_1 = createDataPartition(Rem_DATA$Age, p= 0.5 , list=FALSE)
Train_Data_1 = Rem_DATA[Train_Index_1,] #Training data
Validation_Data_1 = Rem_DATA[-Train_Index_1,] #Validation data
#Data Normalization
train.norm.df_1 <- Train_Data_1</pre>
valid.norm.df_1 <- Validation_Data_1</pre>
test.norm.df_1 <- Test_Data_1</pre>
rem_data.norm.df_1 <- Rem_DATA</pre>
norm.values 1 <- preProcess(Train Data 1[-7], method=c("center", "scale"))
train.norm.df_1[-7] <- predict(norm.values_1, Train_Data_1[-7]) #Training Data
valid.norm.df_1[-7] <- predict(norm.values_1, Validation_Data_1[-7])#Validation Data</pre>
test.norm.df_1[-7] <- predict(norm.values_1, test.norm.df_1[-7]) #Test Data
test.norm.df_1[-7] <- predict(norm.values_1, Test_Data_1[-7])</pre>
rem_data.norm.df_1[-7] <- predict(norm.values_1,Rem_DATA[-7])</pre>
head(test.norm.df_1)
##
             Age
                  Experience
                                 Income
                                           Family
                                                         CCAvg
                                                                Mortgage
## 9 -0.90840439 -0.883582836 0.1435652 0.5333142 -0.780693325 0.4495336
## 28 0.05751618 -0.008054857 1.8189997 -1.2081200
                                                   0.234699617 -0.5532869
## 32 -0.46934959 -0.358266049 -0.9878972 -1.2081200 0.009056741 -0.5532869
## 40 -0.64497151 -0.620924443 0.1218063 1.4040313 -0.724282606 2.1948269
## 42 -0.99621536 -0.971135634 -0.3133715 0.5333142 0.178288898 -0.5532869
## 63 -0.29372767 -0.183160453 -1.1402094 -1.2081200 -0.555050449 -0.5532869
##
     Personal.Loan Securities.Account CD.Account
                                                   Online CreditCard
## 9
                 0
                          ## 28
                 0
## 32
                0
                          ## 40
                 0
                          ## 42
                 0
                          -0.3360202 -0.2646808 -1.1857637 -0.6350646
## 63
                 0
                          -0.3360202 -0.2646808 -1.1857637 -0.6350646
     Education_1 Education_2 Education_3
##
## 9
       -0.827392 -0.6607293
                              1.566207
        1.208013 -0.6607293
## 28
                              -0.638166
## 32
       -0.827392 -0.6607293
                               1.566207
## 40
       -0.827392
                 1.5127224
                              -0.638166
## 42
        1.208013 -0.6607293
                              -0.638166
## 63
        1.208013 -0.6607293
                              -0.638166
\#Perfoming\ k-NN\ classification\ on\ Training\ Data,\ k=1
set.seed(1234)
prediction_Q5 <- knn(train = train.norm.df_1[,-7], test = valid.norm.df_1[,-7],</pre>
         cl = train.norm.df_1[,7], k = 1, prob=TRUE)
actual= valid.norm.df_1$Personal.Loan
prediction_prob = attr(prediction_Q5,"prob")
table(prediction Q5,actual) #confusion matrix for the best k value =1
##
               actual
## prediction Q5
                  0
                       1
```

```
##
               0 1795 69
##
                 16 119
               1
mean(prediction_Q5==actual) #accuracy of the best k=1
## [1] 0.9574787
set.seed(1234)
prediction_Q5 <- knn(train = rem_data.norm.df_1[,-7], test = test.norm.df_1[,-7],</pre>
          cl = rem_data.norm.df_1[,7], k = 1, prob=TRUE)
actual= test.norm.df_1$Personal.Loan
prediction_prob = attr(prediction_Q5,"prob")
table(prediction_Q5,actual) #confusion matrix for the best k value =1
##
                actual
## prediction_Q5     0     1
##
              0 907 25
##
              1 12 57
```