

# Assignment 2

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2/17/2022

```
#First load the all the required packages using library for smooth execution.
```

```
library('caret')
```

```
## Loading required package: ggplot2
```

```
## Loading required package: lattice
```

```
library('ISLR')
```

```
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
```

```
library('class')
```

```
library(readr)
```

```
library(gmodels)
```

```
library(FNN)
```

```
##
```

```
## Attaching package: 'FNN'
```

```
## The following objects are masked from 'package:class':
```

```
##
```

```
##      knn, knn.cv
```

```
#To import the UniversalBank.csv file
```

```
Universal_Bank <- read.csv("C:/Users/lavak/Documents/R/Assignment2/UniversalBank.csv")
```

```
colnames(Universal_Bank) #for displaying the column names
```

```
## [1] "ID" "Age" "Experience"
## [4] "Income" "ZIP.Code" "Family"
## [7] "CCAvg" "Education" "Mortgage"
## [10] "Personal.Loan" "Securities.Account" "CD.Account"
## [13] "Online" "CreditCard"
```

```
summary(Universal_Bank)
```

```
##      ID      Age      Experience      Income      ZIP.Code
## Min.   : 1    Min.   :23.00    Min.   : -3.0    Min.   : 8.00    Min.   : 9307
## 1st Qu.:1251  1st Qu.:35.00    1st Qu.:10.0    1st Qu.: 39.00    1st Qu.:91911
## Median :2500  Median :45.00    Median :20.0    Median : 64.00    Median :93437
## Mean   :2500  Mean   :45.34    Mean   :20.1    Mean   : 73.77    Mean   :93153
## 3rd Qu.:3750  3rd Qu.:55.00    3rd Qu.:30.0    3rd Qu.: 98.00    3rd Qu.:94608
## Max.   :5000  Max.   :67.00    Max.   :43.0    Max.   :224.00    Max.   :96651
##      Family      CCAvg      Education      Mortgage
## Min.   :1.000    Min.   : 0.000    Min.   :1.000    Min.   : 0.0
## 1st Qu.:1.000    1st Qu.: 0.700    1st Qu.:1.000    1st Qu.: 0.0
## Median :2.000    Median : 1.500    Median :2.000    Median : 0.0
## Mean   :2.396    Mean   : 1.938    Mean   :1.881    Mean   : 56.5
## 3rd Qu.:3.000    3rd Qu.: 2.500    3rd Qu.:3.000    3rd Qu.:101.0
## Max.   :4.000    Max.   :10.000    Max.   :3.000    Max.   :635.0
## Personal.Loan  Securities.Account  CD.Account      Online
## Min.   :0.000    Min.   :0.0000    Min.   :0.0000    Min.   :0.0000
## 1st Qu.:0.000    1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000
## Median :0.000    Median :0.0000    Median :0.0000    Median :1.0000
## Mean   :0.096    Mean   :0.1044    Mean   :0.0604    Mean   :0.5968
## 3rd Qu.:0.000    3rd Qu.:0.0000    3rd Qu.:0.0000    3rd Qu.:1.0000
## Max.   :1.000    Max.   :1.0000    Max.   :1.0000    Max.   :1.0000
##      CreditCard
## Min.   :0.000
## 1st Qu.:0.000
## Median :0.000
## Mean   :0.294
## 3rd Qu.:1.000
## Max.   :1.000
```

*#Removing columns ID and ZIP.Code as per the Question instruction by assigning them to NULL*

```
Universal_Bank$ID <- NULL
Universal_Bank$ZIP.Code <- NULL
summary(Universal_Bank)
```

```
##      Age      Experience      Income      Family
## Min.   :23.00    Min.   : -3.0    Min.   : 8.00    Min.   :1.000
## 1st Qu.:35.00    1st Qu.:10.0    1st Qu.: 39.00    1st Qu.:1.000
## Median :45.00    Median :20.0    Median : 64.00    Median :2.000
## Mean   :45.34    Mean   :20.1    Mean   : 73.77    Mean   :2.396
## 3rd Qu.:55.00    3rd Qu.:30.0    3rd Qu.: 98.00    3rd Qu.:3.000
## Max.   :67.00    Max.   :43.0    Max.   :224.00    Max.   :4.000
##      CCAvg      Education      Mortgage      Personal.Loan
## Min.   : 0.000    Min.   :1.000    Min.   : 0.0    Min.   :0.000
## 1st Qu.: 0.700    1st Qu.:1.000    1st Qu.: 0.0    1st Qu.:0.000
## Median : 1.500    Median :2.000    Median : 0.0    Median :0.000
```

```
## Mean : 1.938 Mean :1.881 Mean : 56.5 Mean :0.096
## 3rd Qu.: 2.500 3rd Qu.:3.000 3rd Qu.:101.0 3rd Qu.:0.000
## Max. :10.000 Max. :3.000 Max. :635.0 Max. :1.000
## Securities.Account CD.Account Online CreditCard
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.000
## Median :0.0000 Median :0.0000 Median :1.0000 Median :0.000
## Mean :0.1044 Mean :0.0604 Mean :0.5968 Mean :0.294
## 3rd Qu.:0.0000 3rd Qu.:0.0000 3rd Qu.:1.0000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.000
```

*#Making the Personal Loan column as factor value*

```
Universal_Bank$Personal.Loan = as.factor(Universal_Bank$Personal.Loan)
```

*#Applying normalization to Universal\_Bank dataset*

```
Normal_model <- preProcess(Universal_Bank,method = "range")
Universal_Bank_Norm <- predict(Normal_model,Universal_Bank)
summary(Universal_Bank_Norm)
```

```
## Age Experience Income Family
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.2727 1st Qu.:0.2826 1st Qu.:0.1435 1st Qu.:0.0000
## Median :0.5000 Median :0.5000 Median :0.2593 Median :0.3333
## Mean :0.5077 Mean :0.5023 Mean :0.3045 Mean :0.4655
## 3rd Qu.:0.7273 3rd Qu.:0.7174 3rd Qu.:0.4167 3rd Qu.:0.6667
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.0000
## CCAvg Education Mortgage Personal.Loan
## Min. :0.0000 Min. :0.0000 Min. :0.00000 0:4520
## 1st Qu.:0.0700 1st Qu.:0.0000 1st Qu.:0.00000 1: 480
## Median :0.1500 Median :0.5000 Median :0.00000
## Mean :0.1938 Mean :0.4405 Mean :0.08897
## 3rd Qu.:0.2500 3rd Qu.:1.0000 3rd Qu.:0.15906
## Max. :1.0000 Max. :1.0000 Max. :1.00000
## Securities.Account CD.Account Online CreditCard
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.000
## Median :0.0000 Median :0.0000 Median :1.0000 Median :0.000
## Mean :0.1044 Mean :0.0604 Mean :0.5968 Mean :0.294
## 3rd Qu.:0.0000 3rd Qu.:0.0000 3rd Qu.:1.0000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.000
```

*#As per the question instructions partitioning the data into 60% for training and 40% for testing*

```
Train_index <- createDataPartition(Universal_Bank$Personal.Loan, p = 0.6, list = FALSE)
train.df = Universal_Bank_Norm[Train_index,]
validation.df = Universal_Bank_Norm[-Train_index,]
```

*#Task 1 classifying the customer as per the data provided in the question 1*

```
To_Predict = data.frame(Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2, Education = 1, Mortgage = 0, Securities.Account = 0)
print(To_Predict)
```

```
## Age Experience Income Family CCAvg Education Mortgage Securities.Account
```

```
## 1 40      10      84      2      2      1      0      0
##   CD.Account Online CreditCard
## 1      0      1      1
```

```
Prediction <- knn(train = train.df[,1:7],
                  test = To_Predict[,1:7], cl = train.df$Personal.Loan, k = 1)
print(Prediction)
```

```
## [1] 1
## attr(,"nn.index")
##      [,1]
## [1,] 1804
## attr(,"nn.dist")
##      [,1]
## [1,] 92.34726
## Levels: 1
```

```
#Customer is classified as 1.
```

### *#Task2*

```
set.seed(123)
Universal_Bank_control <- trainControl(method= "repeatedcv", number = 3, repeats = 2)
searchGrid = expand.grid(k=1:10)

knn.model = train(Personal.Loan~., data = train.df, method = 'knn', tuneGrid = searchGrid, trControl = Universal_Bank_control)
knn.model
```

```
## k-Nearest Neighbors
##
## 3000 samples
## 11 predictor
## 2 classes: '0', '1'
##
## No pre-processing
## Resampling: Cross-Validated (3 fold, repeated 2 times)
## Summary of sample sizes: 2000, 2000, 2000, 2000, 2000, 2000, ...
## Resampling results across tuning parameters:
##
##   k  Accuracy  Kappa
##   1  0.9546667  0.7070340
##   2  0.9460000  0.6543675
##   3  0.9533333  0.6759682
##   4  0.9490000  0.6382196
##   5  0.9458333  0.6016805
##   6  0.9428333  0.5783428
##   7  0.9396667  0.5421195
##   8  0.9385000  0.5310286
##   9  0.9375000  0.5165528
##  10  0.9358333  0.4984294
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was k = 1.
```

*#The choice of K that balances between overfitting and ignoring predictor information appears as K=3*

*#Question 3*

*#confusion matrix for the validation data that results from using the best k.*

```
predictions <- predict(knn.model,validation.df)

confusionMatrix(predictions,validation.df$Personal.Loan)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1789   54
##           1   19  138
##
##           Accuracy : 0.9635
##           95% CI : (0.9543, 0.9713)
##       No Information Rate : 0.904
##       P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.7711
##
##  Mcnemar's Test P-Value : 6.909e-05
##
##           Sensitivity : 0.9895
##           Specificity : 0.7188
##           Pos Pred Value : 0.9707
##           Neg Pred Value : 0.8790
##           Prevalence : 0.9040
##           Detection Rate : 0.8945
##       Detection Prevalence : 0.9215
##           Balanced Accuracy : 0.8541
##
##           'Positive' Class : 0
##
```

*#Question 4*

*#Classify the customer using the best k*

```
To_Predict_Normaliz = data.frame(Age = 40, Experience = 10, Income = 84, Family = 2,
                                CCAvg = 2, Education = 1, Mortgage = 0,
                                Securities.Account = 0, CD.Account = 0, Online = 1,
                                CreditCard = 1)
To_Predict_Normaliz = predict(Normal_model, To_Predict)
predict(knn.model, To_Predict_Normaliz)
```

```
## [1] 0
## Levels: 0 1
```

*#Question 5*

*#As per the question instructions Repartition the data into 50% for training ,30% for validation, 20% for test*

```
train_size = 0.5
```

```

Train_index = createDataPartition(Universal_Bank$Personal.Loan, p = 0.5, list = FALSE)
train.df = Universal_Bank_Norm[Train_index,]

test_size = 0.2
Test_index = createDataPartition(Universal_Bank$Personal.Loan, p = 0.2, list = FALSE)
Test.df = Universal_Bank_Norm[Test_index,]

valid_size = 0.3
Validation_index = createDataPartition(Universal_Bank$Personal.Loan, p = 0.3, list = FALSE)
validation.df = Universal_Bank_Norm[Validation_index,]

Testingknn <- knn(train = train.df[, -8], test = Test.df[, -8], cl = train.df[, 8], k = 3)
Validationknn <- knn(train = train.df[, -8], test = validation.df[, -8], cl = train.df[, 8], k = 3)
Trainingknn <- knn(train = train.df[, -8], test = train.df[, -8], cl = train.df[, 8], k = 3)

#Comparing the confusion matrix of the test set with that of the training and validation sets.

confusionMatrix(Testingknn, Test.df[, 8])

```

```

## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 901  36
##           1   3  60
##
##           Accuracy : 0.961
##           95% CI : (0.9471, 0.9721)
##       No Information Rate : 0.904
##       P-Value [Acc > NIR] : 5.695e-12
##
##           Kappa : 0.7345
##
##  Mcnemar's Test P-Value : 2.990e-07
##
##           Sensitivity : 0.9967
##           Specificity : 0.6250
##       Pos Pred Value : 0.9616
##       Neg Pred Value : 0.9524
##           Prevalence : 0.9040
##       Detection Rate : 0.9010
##   Detection Prevalence : 0.9370
##       Balanced Accuracy : 0.8108
##
##       'Positive' Class : 0
##

```

```

confusionMatrix(Trainingknn, train.df[, 8])

```

```

## Confusion Matrix and Statistics
##
##           Reference

```

```

## Prediction    0    1
##           0 2254   52
##           1    6  188
##
##           Accuracy : 0.9768
##           95% CI : (0.9701, 0.9823)
##           No Information Rate : 0.904
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.8538
##
## Mcnemar's Test P-Value : 3.446e-09
##
##           Sensitivity : 0.9973
##           Specificity : 0.7833
##           Pos Pred Value : 0.9775
##           Neg Pred Value : 0.9691
##           Prevalence : 0.9040
##           Detection Rate : 0.9016
##           Detection Prevalence : 0.9224
##           Balanced Accuracy : 0.8903
##
##           'Positive' Class : 0
##

```

```

confusionMatrix(Validationknn, validation.df[,8])

```

```

## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1347   45
##           1    9   99
##
##           Accuracy : 0.964
##           95% CI : (0.9533, 0.9728)
##           No Information Rate : 0.904
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.7665
##
## Mcnemar's Test P-Value : 1.908e-06
##
##           Sensitivity : 0.9934
##           Specificity : 0.6875
##           Pos Pred Value : 0.9677
##           Neg Pred Value : 0.9167
##           Prevalence : 0.9040
##           Detection Rate : 0.8980
##           Detection Prevalence : 0.9280
##           Balanced Accuracy : 0.8404
##
##           'Positive' Class : 0
##

```

*#After comparing the data obtained from both confusion matrices. We can observe that training accuracy is*  
*#We can also determine that Training accuracy is slightly higher than the test and validation sets which*