

Assignment 2

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2/19/2022

R Markdown

```
UniversalBank <- read.csv("~/Downloads/UniversalBank.csv")
summary(UniversalBank)
```

```
##      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
:93152
## 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
```

#Null Variables

```
UniversalBank$ID<-NULL
```

```
UniversalBank$ZIP.Code<-NULL
```

```
summary(UniversalBank)
```

```
##      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
```

#CaretLibrary

```
library(caret)
```

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

```
## Warning in register(): Can't find generic `scale_type` in package ggplot2
to
```

```
## register S3 method.
```

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

```
library(class)
```

```
summary(UniversalBank)
```

```
##      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
```

#DummyVariable

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

```
summary(UniversalBank)
```

```
## 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 0:4520
## 1st Qu.: 0.700 1st Qu.:1.000 1st Qu.: 0.0 1: 480
## Median : 1.500 Median :2.000 Median : 0.0
## Mean : 1.938 Mean :1.881 Mean : 56.5
## 3rd Qu.: 2.500 3rd Qu.:3.000 3rd Qu.:101.0
## Max. :10.000 Max. :3.000 Max. :635.0
## 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
```

```
Bank_Norm<-UniversalBank
```

#Normalize

```
Norm_model<-preProcess(UniversalBank[,-8],method = c("center","scale"))
Bank_Norm[,-8]=predict(Norm_model,UniversalBank[,-8])
summary(Bank_Norm)
```

```
##      Age      Experience      Income      Family
## Min.    :-1.94871  Min.    :-2.014710  Min.    :-1.4288  Min.    :-1.2167
## 1st Qu.: -0.90188  1st Qu.: -0.881116  1st Qu.: -0.7554  1st Qu.: -1.2167
## Median :-0.02952  Median :-0.009121  Median :-0.2123  Median :-0.3454
## Mean   : 0.00000  Mean   : 0.000000  Mean   : 0.0000  Mean   : 0.0000
## 3rd Qu.: 0.84284  3rd Qu.: 0.862874  3rd Qu.: 0.5263  3rd Qu.: 0.5259
## Max.    : 1.88967  Max.    : 1.996468  Max.    : 3.2634  Max.    : 1.3973
##      CCAvg      Education      Mortgage      Personal.Loan
## Min.    :-1.1089  Min.    :-1.0490  Min.    :-0.5555  0:4520
## 1st Qu.: -0.7083  1st Qu.: -1.0490  1st Qu.: -0.5555  1: 480
## Median :-0.2506  Median : 0.1417  Median :-0.5555
## Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.0000
## 3rd Qu.: 0.3216  3rd Qu.: 1.3324  3rd Qu.: 0.4375
## Max.    : 4.6131  Max.    : 1.3324  Max.    : 5.6875
## Securities.Account  CD.Account      Online      CreditCard
## Min.    :-0.3414  Min.    :-0.2535  Min.    :-1.2165  Min.    :-0.6452
## 1st Qu.: -0.3414  1st Qu.: -0.2535  1st Qu.: -1.2165  1st Qu.: -0.6452
## Median :-0.3414  Median :-0.2535  Median : 0.8219  Median :-0.6452
## Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.0000
## 3rd Qu.: -0.3414  3rd Qu.: -0.2535  3rd Qu.: 0.8219  3rd Qu.: 1.5495
## Max.    : 2.9286  Max.    : 3.9438  Max.    : 0.8219  Max.    : 1.5495
```

#Train

```
Train_Index=createDataPartition(UniversalBank$Personal.Loan,p=0.6,list =
FALSE)
Train.df=Bank_Norm[Train_Index,]
Validation.df=Bank_Norm[-Train_Index,]
```

#Predict

```
To_Predict=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)
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

To_Predict_norm<-predict(Norm_model,To_Predict)
```

#Prediction

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

print(Prediction)
```

```

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

#Task2

set.seed(123)
fitControl<-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 = fitControl)

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  0.9548333  0.7119388
##  2  0.9496667  0.6789197
##  3  0.9556667  0.7018490
##  4  0.9526667  0.6751525
##  5  0.9538333  0.6786610
##  6  0.9540000  0.6799036
##  7  0.9513333  0.6543835
##  8  0.9491667  0.6360260
##  9  0.9480000  0.6230283
## 10  0.9465000  0.6079954
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was k = 3.

predictions<-predict(Knn.model,Validation.df)

confusionMatrix(predictions,Validation.df$Personal.Loan)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    0      1
##      0 1804    76

```

```

##          1      4   116
##
##          Accuracy : 0.96
##          95% CI : (0.9505, 0.9682)
##    No Information Rate : 0.904
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.7231
##
##  McNemar's Test P-Value : 2.054e-15
##
##          Sensitivity : 0.9978
##          Specificity : 0.6042
##          Pos Pred Value : 0.9596
##          Neg Pred Value : 0.9667
##          Prevalence : 0.9040
##          Detection Rate : 0.9020
##    Detection Prevalence : 0.9400
##          Balanced Accuracy : 0.8010
##
##          'Positive' Class : 0
##

To_Predict_norm=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_norm=predict(Norm_model,To_Predict)
predict(Knn.model,To_Predict_norm)

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

#Task 5

train_size=.5
Train_Index=createDataPartition(UniversalBank$Personal.Loan,p=0.5,list =
FALSE)
Train.df=Bank_Norm[Train_Index,]
valid_size=.30
Validation.df=Bank_Norm[-Train_Index,]

To_Predict=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)
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

To_Predict_norm<-predict(Norm_model,To_Predict)

#Prediction
Prediction <-knn(train=Train.df[,1:7],
                 test=To_Predict_norm[,1:7],
                 cl=Train.df$Personal.Loan,
                 k=1)

print(Prediction)

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

set.seed(123)
fitControl<-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 = fitControl)

Knn.model

## k-Nearest Neighbors
##
## 2500 samples
## 11 predictor
## 2 classes: '0', '1'
##
## No pre-processing
## Resampling: Cross-Validated (3 fold, repeated 2 times)
## Summary of sample sizes: 1667, 1667, 1666, 1667, 1666, 1667, ...
## Resampling results across tuning parameters:
##
## k Accuracy Kappa
## 1 0.9544007 0.7026621
## 2 0.9455996 0.6436537
## 3 0.9530009 0.6745506
## 4 0.9514005 0.6552616
## 5 0.9514000 0.6501013
## 6 0.9478000 0.6214736
## 7 0.9443991 0.5838148
## 8 0.9443998 0.5823568
## 9 0.9406005 0.5430711
## 10 0.9404009 0.5414687
##

```

```
## Accuracy was used to select the optimal model using the largest value.  
## The final value used for the model was k = 1.
```

```
predictions<-predict(Knn.model,Validation.df)
```

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

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
```

```
## Prediction    0    1
```

```
##           0 2233   87
```

```
##           1   27  153
```

```
##
```

```
##           Accuracy : 0.9544
```

```
##           95% CI : (0.9455, 0.9622)
```

```
## No Information Rate : 0.904
```

```
## P-Value [Acc > NIR] : < 2.2e-16
```

```
##
```

```
##           Kappa : 0.7042
```

```
##
```

```
## McNemar's Test P-Value : 3.279e-08
```

```
##
```

```
##           Sensitivity : 0.9881
```

```
##           Specificity : 0.6375
```

```
## Pos Pred Value : 0.9625
```

```
## Neg Pred Value : 0.8500
```

```
## Prevalence : 0.9040
```

```
## Detection Rate : 0.8932
```

```
## Detection Prevalence : 0.9280
```

```
## Balanced Accuracy : 0.8128
```

```
##
```

```
## 'Positive' Class : 0
```

```
##
```

#The Difference that I noticed was the increase in accuracy from 94.6% in the first set and up to 95% when calculating the confusion matrix. #This is because training and validating in smaller groups improves models and allows them to perform more accurately.