

Assignment 2

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```
library(caret)
```

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

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

```
library(lattice)
```

```
library(class)
```

```
library(readr)
```

```
universalbank <- read.csv("universalbank.csv")
```

```
colnames(universalbank) <- c('ID', 'Age', 'Experience', 'Income', 'ZIP Code', 'Family', 'CCAvg', 'Education',  
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   :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
```

Question 1: Use "Null" function to remove variables not included in the mode, then transform character

```
universalbank$ID <- NULL
universalbank$`ZIP Code` <- NULL
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
```

Normalize the whole data before running the model.

```
Norm_model <- preProcess(universalbank, method = c("center", "scale"))
universalbank_norm=predict(Norm_model,universalbank)
summary(universalbank_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
```

```
universalbank_norm$Personal.Loan = universalbank$Personal.Loan
```

```
# Partition the data
train_index = createDataPartition(universalbank$Personal.Loan,p=0.6, list=FALSE)
train.df=universalbank_norm[train_index,]
Validation.df=universalbank_norm[-train_index,]
# Create the test data
To_Predict=data.frame(Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2,
                      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)
print(To_Predict_norm)
```

```
##           Age Experience      Income      Family      CCAvg Education Mortgage
## 1 -0.4657003 -0.8811162 0.2221371 -0.3453975 0.0355115 -1.048973 -0.5554684
##   Securities.Account CD.Account      Online CreditCard
## 1           -0.3413892 -0.2535149 0.8218687  1.549477
```

```
Prediction <-knn(train=train.df[,1:7,9:12],
                 test=To_Predict_norm[,1:7,9:12],
                 cl=train.df$Personal.Loan,
                 k=1)
```

```
## Warning in drop && !has.j: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && length(y) == 1L: 'length(x) = 4 > 1' in coercion to
## 'logical(1)'
```

```
## Warning in drop && !mdrop: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && !has.j: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && length(y) == 1L: 'length(x) = 4 > 1' in coercion to
## 'logical(1)'
```

```
## Warning in drop && !mdrop: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
print(Prediction)
```

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

```
#Question 2 k=1 got highest accuracy of 0.953.  
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 0.9505000 0.6784999  
## 2 0.9455000 0.6475715  
## 3 0.9531667 0.6806574  
## 4 0.9530000 0.6738620  
## 5 0.9533333 0.6730544  
## 6 0.9508333 0.6511946  
## 7 0.9501667 0.6435187  
## 8 0.9478333 0.6227036  
## 9 0.9488333 0.6287298  
## 10 0.9491667 0.6275120  
##  
## Accuracy was used to select the optimal model using the largest value.  
## The final value used for the model was k = 5.
```

```
#Question3 Accuracy recoded as 0.958.  
predictions <- predict(Knn.model, Validation.df)  
confusionMatrix(predictions, Validation.df$Personal.Loan)
```

```
## Confusion Matrix and Statistics  
##  
##           Reference  
## Prediction    0    1  
##           0 1801   85
```

```
##          1      7   107
##
##          Accuracy : 0.954
##          95% CI : (0.9439, 0.9628)
##    No Information Rate : 0.904
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.6762
##
## Mcnemar's Test P-Value : 9.923e-16
##
##          Sensitivity : 0.9961
##          Specificity : 0.5573
##    Pos Pred Value : 0.9549
##    Neg Pred Value : 0.9386
##          Prevalence : 0.9040
##    Detection Rate : 0.9005
##    Detection Prevalence : 0.9430
##    Balanced Accuracy : 0.7767
##
##    'Positive' Class : 0
##
```

#Question 4

```
To_Predict = data.frame(Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2,
                          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
```

#Question 5

```
splitSample <- sample(1:3, size=nrow(universalbank_norm), prob=c(0.5,0.3,0.2), replace = TRUE)
train_data <- universalbank_norm[splitSample==1,]
valid_data <- universalbank_norm[splitSample==2,]
test_data <- universalbank_norm[splitSample==3,]
Predict = data.frame(Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2,
                     CD.Account = 0, Online = 1, CreditCard = 1)
print(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
```

```
Predict_norm <- predict(Norm_model, Predict)
print(Predict_norm)
```

```
##           Age Experience      Income      Family      CCAvg Education Mortgage
## 1 -0.4657003 -0.8811162 0.2221371 -0.3453975 0.0355115 -1.048973 -0.5554684
##   Securities.Account CD.Account      Online CreditCard
## 1          -0.3413892 -0.2535149 0.8218687  1.549477
```

```
Prediction_newsplit <-knn(train=train.df[,1:7,9:12],
                           test=To_Predict_norm[,1:7,9:12],
                           cl=train.df$Personal.Loan,
                           k=1)
```

```
## Warning in drop && !has.j: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && length(y) == 1L: 'length(x) = 4 > 1' in coercion to
## 'logical(1)'
```

```
## Warning in drop && !mdrop: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && !has.j: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
## Warning in drop && length(y) == 1L: 'length(x) = 4 > 1' in coercion to
## 'logical(1)'
```

```
## Warning in drop && !mdrop: 'length(x) = 4 > 1' in coercion to 'logical(1)'
```

```
print(Prediction_newsplit)
```

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

```
fitControl2 <- trainControl(method = "repeatedcv",
                             number = 3,
                             repeats = 2)
                             searchGrid=expand.grid(k = 1:10)
Knn.model2 = train(Personal.Loan~.,
                    data=train.df,
                    method='knn',
                    tuneGrid=searchGrid,
                    trControl = fitControl2,)
Knn.model2
```

```
## 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 0.9528333 0.6878298
## 2 0.9460000 0.6539164
## 3 0.9531667 0.6815989
```

```
## 4 0.9520000 0.6715692
## 5 0.9528333 0.6676317
## 6 0.9511667 0.6536142
## 7 0.9508333 0.6458572
## 8 0.9481667 0.6223505
## 9 0.9475000 0.6155112
## 10 0.9453333 0.5947697
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was k = 3.
```

```
predictions2 <- predict(Knn.model2, Validation.df)
confusionMatrix(predictions2, Validation.df$Personal.Loan)
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1796   75
##           1   12  117
##
##           Accuracy : 0.9565
##           95% CI : (0.9466, 0.965)
##           No Information Rate : 0.904
##           P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.7063
##
## Mcnemar's Test P-Value : 2.989e-11
##
##           Sensitivity : 0.9934
##           Specificity : 0.6094
##           Pos Pred Value : 0.9599
##           Neg Pred Value : 0.9070
##           Prevalence : 0.9040
##           Detection Rate : 0.8980
##           Detection Prevalence : 0.9355
##           Balanced Accuracy : 0.8014
##
##           'Positive' Class : 0
##
```