

```
#install.packages
library(caret)
library(ISLR)
library(gmodels)
library(ggplot2)
library(FNN)
library(dummies)
library(fastDummies)
library(e1071)
library(dplyr)

#install.packages("caret")
library(caret)
# install.packages("ISLR") # only install if needed
library(ISLR)
```

```
library(readxl)
uk <- read.csv("C:/Users/mnooo/Desktop/Datasets/UniversalBank.csv")
#View(uk)
```

```
# Clean up data and get it ready for KNN
library(dplyr)
```

```
uk$Education_1 = ifelse(uk$Education ==1, 1, 0)
uk$Education_2 = ifelse(uk$Education ==2, 1, 0)
uk$Education_3 = ifelse(uk$Education ==3, 1, 0)
uk$Personal_Loan = as.factor(uk$Personal.Loan)
```

```
View(uk)
```

```
uk1 <- uk[, c(2:4, 6:7, 9, 11:18)] # Select a subset of variables, without ID and zipcode, original variable Personal.Loan
head(uk1)
```

```
##   Age Experience Income Family CCAvg Mortgage Securities.Account CD.Account
## 1  25          1     49      4   1.6         0              1          0
## 2  45         19     34      3   1.5         0              1          0
## 3  39         15     11      1   1.0         0              0          0
## 4  35          9    100      1   2.7         0              0          0
## 5  35          8     45      4   1.0         0              0          0
## 6  37         13     29      4   0.4        155              0          0
##   Online CreditCard Education_1 Education_2 Education_3 Personal_Loan
## 1      0           0            1           0           0           0
## 2      0           0            1           0           0           0
## 3      0           0            1           0           0           0
## 4      0           0            0           1           0           0
## 5      0           1            0           1           0           0
## 6      1           0            0           1           0           0
```

```
str(uk1)
```

```
## 'data.frame':    5000 obs. of  14 variables:
## $ Age           : int  25 45 39 35 35 37 53 50 35 34 ...
## $ Experience     : int  1 19 15 9 8 13 27 24 10 9 ...
## $ Income         : int  49 34 11 100 45 29 72 22 81 180 ...
## $ Family         : int  4 3 1 1 4 4 2 1 3 1 ...
## $ CCAvg          : num  1.6 1.5 1 2.7 1 0.4 1.5 0.3 0.6 8.9 ...
## $ Mortgage       : int  0 0 0 0 0 155 0 0 104 0 ...
## $ Securities.Account: int  1 1 0 0 0 0 0 0 0 0 ...
## $ CD.Account      : int  0 0 0 0 0 0 0 0 0 0 ...
## $ Online          : int  0 0 0 0 0 1 1 0 1 0 ...
## $ CreditCard      : int  0 0 0 0 1 0 0 1 0 0 ...
## $ Education_1     : num  1 1 1 0 0 0 0 0 0 0 ...
## $ Education_2     : num  0 0 0 1 1 1 1 0 1 0 ...
## $ Education_3     : num  0 0 0 0 0 0 0 1 0 1 ...
## $ Personal_Loan   : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 1 1 2 ...
```

```
set.seed(15)
Test_Index = createDataPartition(uk1$Age,p=0.2, list=FALSE) # 20% reserved for Test
Test_Data = uk1[Test_Index,]
TraVal_Data = uk1[-Test_Index,] # Validation and Training data is rest
Train_Index = createDataPartition(TraVal_Data$Age,p=0.75, list=FALSE) # 75% of remaining data as training

# 75% of the 80% is 60% which is training

Train_Data = TraVal_Data[Train_Index,]
Validation_Data = TraVal_Data[-Train_Index,] # rest as validation
View(Train_Data)
summary(Train_Data)
```

```
##      Age      Experience      Income      Family
## Min.   :23.00   Min.    :-3.00   Min.    : 8.0   Min.    :1.000
## 1st Qu.:35.00   1st Qu.:10.00   1st Qu.: 38.0   1st Qu.:1.000
## Median :45.00   Median :20.00   Median : 63.0   Median :2.000
## Mean   :45.37   Mean    :20.13   Mean    : 72.3   Mean    :2.401
## 3rd Qu.:55.00   3rd Qu.:30.00   3rd Qu.: 95.0   3rd Qu.:3.000
## Max.    :67.00   Max.    :43.00   Max.    :218.0   Max.    :4.000
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.    :0.0000   Min.     : 0.00   Min.    :0.0000   Min.    :0.00000
## 1st Qu.:0.6925   1st Qu.: 0.00   1st Qu.:0.0000   1st Qu.:0.00000
## Median :1.5000   Median : 0.00   Median :0.0000   Median :0.00000
## Mean    :1.8996   Mean     :55.77   Mean    :0.1037   Mean    :0.06033
## 3rd Qu.:2.5000   3rd Qu.:98.00   3rd Qu.:0.0000   3rd Qu.:0.00000
## Max.    :9.3000   Max.    :635.00   Max.    :1.0000   Max.    :1.00000
##      Online      CreditCard      Education_1      Education_2
## Min.    :0.000   Min.    :0.000   Min.    :0.000   Min.    :0.0000
## 1st Qu.:0.000   1st Qu.:0.000   1st Qu.:0.000   1st Qu.:0.0000
## Median :1.000   Median :0.000   Median :0.000   Median :0.0000
## Mean    :0.599   Mean     :0.301   Mean    :0.405   Mean    :0.2887
## 3rd Qu.:1.000   3rd Qu.:1.000   3rd Qu.:1.000   3rd Qu.:1.0000
## Max.    :1.000   Max.    :1.000   Max.    :1.000   Max.    :1.0000
##      Education_3      Personal_Loan
## Min.    :0.0000      0:2718
## 1st Qu.:0.0000      1: 282
## Median :0.0000
## Mean    :0.3063
## 3rd Qu.:1.0000
## Max.    :1.0000
```

```
summary(Validation_Data)
```

```
##      Age      Experience      Income      Family
## Min.   :23.00   Min.    :-2.00   Min.    : 8.00   Min.    :1.000
## 1st Qu.:35.00   1st Qu.:10.00   1st Qu.: 39.00   1st Qu.:1.000
## Median :45.00   Median :20.00   Median : 65.00   Median :2.000
## Mean   :45.19   Mean    :19.92   Mean    : 75.84   Mean    :2.396
## 3rd Qu.:55.00   3rd Qu.:29.00   3rd Qu.:102.50   3rd Qu.:3.000
## Max.    :67.00   Max.     :43.00   Max.     :204.00   Max.     :4.000
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.    : 0.00   Min.     : 0.00   Min.     :0.0000   Min.     :0.00000
## 1st Qu.: 0.70   1st Qu.: 0.00   1st Qu.:0.0000   1st Qu.:0.00000
## Median : 1.60   Median : 0.00   Median :0.0000   Median :0.00000
## Mean    : 2.03   Mean     :59.97   Mean     :0.0951   Mean     :0.06006
## 3rd Qu.: 2.60   3rd Qu.:110.00   3rd Qu.:0.0000   3rd Qu.:0.00000
## Max.    :10.00   Max.     :617.00   Max.     :1.0000   Max.     :1.00000
##      Online      CreditCard      Education_1      Education_2
## Min.     :0.0000   Min.     :0.0000   Min.     :0.0000   Min.     :0.0000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000
## Median :1.0000   Median :0.0000   Median :0.0000   Median :0.0000
## Mean    :0.5696   Mean     :0.2803   Mean     :0.4454   Mean     :0.2633
## 3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:1.0000
## Max.     :1.0000   Max.     :1.0000   Max.     :1.0000   Max.     :1.0000
##      Education_3      Personal_Loan
## Min.     :0.0000      0:902
## 1st Qu.:0.0000      1: 97
## Median :0.0000
## Mean     :0.2913
## 3rd Qu.:1.0000
## Max.     :1.0000
```

```
summary(Test_Data)
```

```
##      Age      Experience      Income      Family
## Min.   :23.00   Min.    :-3.00   Min.    : 8.00   Min.    :1.000
## 1st Qu.:35.00   1st Qu.:10.00   1st Qu.: 39.00   1st Qu.:1.000
## Median :45.00   Median :20.00   Median : 65.00   Median :2.000
## Mean   :45.39   Mean    :20.21   Mean    : 76.14   Mean    :2.382
## 3rd Qu.:55.00   3rd Qu.:30.00   3rd Qu.:103.00   3rd Qu.:3.000
## Max.    :67.00   Max.     :42.00   Max.     :224.00   Max.     :4.000
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.    : 0.000   Min.     : 0.00   Min.     :0.0000   Min.     :0.00000
## 1st Qu.: 0.700   1st Qu.: 0.00   1st Qu.:0.0000   1st Qu.:0.00000
## Median : 1.600   Median : 0.00   Median :0.0000   Median :0.00000
## Mean    : 1.961   Mean     :55.23   Mean     :0.1159   Mean     :0.06094
## 3rd Qu.: 2.600   3rd Qu.:101.00   3rd Qu.:0.0000   3rd Qu.:0.00000
## Max.    :10.000   Max.     :569.00   Max.     :1.0000   Max.     :1.00000
##      Online      CreditCard      Education_1      Education_2
## Min.     :0.0000   Min.     :0.0000   Min.     :0.0000   Min.     :0.0000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.0000
## Median :1.0000   Median :0.0000   Median :0.0000   Median :0.0000
## Mean    :0.6174   Mean     :0.2867   Mean     :0.4356   Mean     :0.2737
## 3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:1.0000
## Max.     :1.0000   Max.     :1.0000   Max.     :1.0000   Max.     :1.0000
##      Education_3      Personal_Loan
## Min.     :0.0000      0:900
## 1st Qu.:0.0000      1:101
## Median :0.0000
## Mean     :0.2907
## 3rd Qu.:1.0000
## Max.     :1.0000
```

```
str(Train_Data)
```

```
## 'data.frame': 3000 obs. of 14 variables:
## $ Age : int 25 37 53 50 35 34 65 29 48 60 ...
## $ Experience : int 1 13 27 24 10 9 39 5 23 30 ...
## $ Income : int 49 29 72 22 81 180 105 45 114 22 ...
## $ Family : int 4 4 2 1 3 1 4 3 2 1 ...
## $ CCAvg : num 1.6 0.4 1.5 0.3 0.6 8.9 2.4 0.1 3.8 1.5 ...
## $ Mortgage : int 0 155 0 0 104 0 0 0 0 0 ...
## $ Securities.Account: int 1 0 0 0 0 0 0 0 1 0 ...
## $ CD.Account : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Online : int 0 1 1 0 1 0 0 1 0 1 ...
## $ CreditCard : int 0 0 0 1 0 0 0 0 0 1 ...
## $ Education_1 : num 1 0 0 0 0 0 0 0 0 0 ...
## $ Education_2 : num 0 1 1 0 1 0 0 1 0 0 ...
## $ Education_3 : num 0 0 0 1 0 1 1 0 1 1 ...
## $ Personal_Loan : Factor w/ 2 levels "0","1": 1 1 1 1 1 2 1 1 1 1 ...
```

```
str(Validation_Data)
```

```
## 'data.frame':    999 obs. of  14 variables:
## $ Age           : int  35 67 46 46 56 40 48 51 42 57 ...
## $ Experience     : int  8 41 21 20 30 16 24 25 18 32 ...
## $ Income         : int  45 112 193 158 48 29 81 71 141 84 ...
## $ Family         : int  4 1 2 1 1 1 3 1 3 3 ...
## $ CCAvg          : num  1 2 8.1 2.4 2.2 2 0.7 1.4 5 1.6 ...
## $ Mortgage       : int  0 0 0 0 0 0 0 198 0 0 ...
## $ Securities.Account: int  0 1 0 0 0 0 0 0 1 1 ...
## $ CD.Account      : int  0 0 0 0 0 0 0 0 1 0 ...
## $ Online          : int  0 0 0 1 1 1 0 0 1 0 ...
## $ CreditCard      : int  1 0 0 1 1 0 0 0 0 0 ...
## $ Education_1     : num  0 1 0 1 0 0 1 0 0 0 ...
## $ Education_2     : num  1 0 0 0 0 1 0 0 0 0 ...
## $ Education_3     : num  0 0 1 0 1 0 0 1 1 1 ...
## $ Personal_Loan   : Factor w/ 2 levels "0","1": 1 1 2 1 1 1 1 1 2 1 ...
```

```
str(Test_Data)
```

```
## 'data.frame':    1001 obs. of  14 variables:
## $ Age           : int  45 39 35 59 55 57 44 38 59 30 ...
## $ Experience     : int  19 15 9 32 28 27 18 13 35 6 ...
## $ Income         : int  34 11 100 40 21 63 43 119 35 18 ...
## $ Family         : int  3 1 1 4 1 3 2 1 1 3 ...
## $ CCAvg          : num  1.5 1 2.7 2.5 0.5 2 0.7 3.3 1.2 0.9 ...
## $ Mortgage       : int  0 0 0 0 0 0 163 0 122 0 ...
## $ Securities.Account: int  1 0 0 0 1 0 1 0 0 0 ...
## $ CD.Account      : int  0 0 0 0 0 0 0 1 0 0 ...
## $ Online          : int  0 0 0 1 0 1 0 1 1 0 ...
## $ CreditCard      : int  0 0 0 0 1 0 0 1 0 0 ...
## $ Education_1     : num  1 1 0 0 0 0 1 0 0 0 ...
## $ Education_2     : num  0 0 1 1 1 0 0 1 0 0 ...
## $ Education_3     : num  0 0 0 0 0 1 0 0 1 1 ...
## $ Personal_Loan   : Factor w/ 2 levels "0","1": 1 1 1 1 1 1 1 2 1 1 ...
```

```
#Normalization
```

```
# Copy the original data
train.norm.df <- Train_Data
valid.norm.df <- Validation_Data
traval.norm.df <- TraVal_Data
# use preProcess() from the caret package
norm.values <- preProcess(Train_Data[, 1:13], method=c("center", "scale"))
train.norm.df[, 1:13] <- predict(norm.values, Train_Data[, 1:13]) # Replace first two columns with normalized values
valid.norm.df[, 1:13] <- predict(norm.values, Validation_Data[, 1:13])
traval.norm.df[, 1:13] <- predict(norm.values, traval.norm.df[, 1:13])
test.norm.df <- predict(norm.values, Test_Data[, 1:13])
summary(train.norm.df)
```

```
##      Age      Experience      Income      Family
## Min.   :-1.94697  Min.   :-2.01326  Min.   :-1.4254  Min.   :-1.2145
## 1st Qu.: -0.90257  1st Qu.: -0.88168  1st Qu.: -0.7603  1st Qu.: -1.2145
## Median :-0.03223  Median :-0.01123  Median :-0.2061  Median :-0.3478
## Mean   : 0.00000  Mean   : 0.00000  Mean   : 0.0000  Mean   : 0.0000
## 3rd Qu.: 0.83810  3rd Qu.: 0.85922  3rd Qu.: 0.5033  3rd Qu.: 0.5188
## Max.    : 1.88251  Max.    : 1.99080  Max.    : 3.2301  Max.    : 1.3855
##      CCAvg      Mortgage  Securities.Account  CD.Account
## Min.   :-1.1008  Min.   :-0.5545  Min.   :-0.34   Min.   :-0.2533
## 1st Qu.: -0.6995  1st Qu.: -0.5545  1st Qu.: -0.34   1st Qu.: -0.2533
## Median :-0.2316  Median :-0.5545  Median :-0.34   Median :-0.2533
## Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.00    Mean   : 0.0000
## 3rd Qu.: 0.3479  3rd Qu.: 0.4199  3rd Qu.: -0.34   3rd Qu.: -0.2533
## Max.    : 4.2884  Max.    : 5.7589  Max.    : 2.94    Max.    : 3.9458
##      Online      CreditCard  Education_1  Education_2
## Min.   :-1.2220  Min.   :-0.6561  Min.   :-0.8249  Min.   :-0.6369
## 1st Qu.: -1.2220  1st Qu.: -0.6561  1st Qu.: -0.8249  1st Qu.: -0.6369
## Median : 0.8181  Median :-0.6561  Median :-0.8249  Median :-0.6369
## Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.0000  Mean   : 0.0000
## 3rd Qu.: 0.8181  3rd Qu.: 1.5236  3rd Qu.: 1.2119  3rd Qu.: 1.5695
## Max.    : 0.8181  Max.    : 1.5236  Max.    : 1.2119  Max.    : 1.5695
##      Education_3  Personal_Loan
## Min.   :-0.6644  0:2718
## 1st Qu.: -0.6644  1: 282
## Median :-0.6644
## Mean   : 0.0000
## 3rd Qu.: 1.5045
## Max.    : 1.5045
```

```
var(train.norm.df[, 1:13])
```

```

##           Age      Experience      Income      Family
## Age      1.000000000  0.994264125 -0.04770546 -0.070973950
## Experience 0.994264125  1.000000000 -0.03939227 -0.075446235
## Income    -0.047705457 -0.039392266  1.00000000 -0.151397484
## Family    -0.070973950 -0.075446235 -0.15139748  1.000000000
## CCAvg     -0.049316783 -0.047182664  0.65251394 -0.103109735
## Mortgage  -0.013781394 -0.012395885  0.21529220 -0.034950727
## Securities.Account -0.005061513 -0.005532953  0.01228408  0.025768029
## CD.Account  0.003286609  0.005076368  0.17237914 -0.027474975
## Online     0.024303974  0.025353685  0.02449525 -0.036667200
## CreditCard  0.028250379  0.031096133 -0.02285159 -0.006553592
## Education_1 -0.025000360 -0.000752846  0.20725185 -0.105134433
## Education_2 -0.014004551 -0.015287490 -0.12034256  0.122708139
## Education_3  0.040389920  0.015829623 -0.10240570 -0.008665706
##           CCAvg      Mortgage Securities.Account  CD.Account
## Age      -0.049316783 -0.013781394      -0.005061513  0.003286609
## Experience -0.047182664 -0.012395885      -0.005532953  0.005076368
## Income    0.652513941  0.215292195      0.012284083  0.172379144
## Family    -0.103109735 -0.034950727      0.025768029 -0.027474975
## CCAvg     1.000000000  0.120610504      0.022439170  0.131933222
## Mortgage  0.120610504  1.000000000      -0.027071056  0.076415649
## Securities.Account 0.022439170 -0.027071056      1.000000000  0.336345349
## CD.Account  0.131933222  0.076415649      0.336345349  1.000000000
## Online     0.002678438 -0.002206841      0.021667198  0.173046976
## CreditCard  0.002541498 -0.016104656      -0.010992531  0.273216272
## Education_1  0.163927388  0.056909450      0.017921092 -0.020832788
## Education_2 -0.089689490 -0.036875837      0.002955340  0.020857768
## Education_3 -0.086401572 -0.024353833      -0.021989532  0.001681470
##           Online      CreditCard Education_1 Education_2
## Age      0.024303974  0.028250379 -0.025000360 -0.01400455
## Experience 0.025353685  0.031096133 -0.000752846 -0.01528749
## Income    0.024495249 -0.022851594  0.207251847 -0.12034256
## Family    -0.036667200 -0.006553592 -0.105134433  0.12270814
## CCAvg     0.002678438  0.002541498  0.163927388 -0.08968949
## Mortgage  -0.002206841 -0.016104656  0.056909450 -0.03687584
## Securities.Account 0.021667198 -0.010992531  0.017921092  0.00295534
## CD.Account  0.173046976  0.273216272 -0.020832788  0.02085777
## Online     1.000000000 -0.005778322  0.023851417  0.01690939
## CreditCard -0.005778322  1.000000000  0.022627489 -0.01870872
## Education_1  0.023851417  0.022627489  1.000000000 -0.52557024
## Education_2  0.016909392 -0.018708716 -0.525570242  1.00000000
## Education_3 -0.042021915 -0.005705227 -0.548265280 -0.42333430
##           Education_3
## Age      0.040389920
## Experience 0.015829623
## Income    -0.102405696
## Family    -0.008665706
## CCAvg     -0.086401572
## Mortgage  -0.024353833
## Securities.Account -0.021989532
## CD.Account  0.001681470
## Online    -0.042021915
## CreditCard -0.005705227

```



```
## Education_1      -0.548265280
## Education_2      -0.423334296
## Education_3       1.000000000
```

```
summary(valid.norm.df)
```

```
##      Age      Experience      Income      Family
## Min.   :-1.94697  Min.   :-1.92621  Min.   :-1.42542  Min.   :-1.214482
## 1st Qu.: -0.90257  1st Qu.: -0.88168  1st Qu.: -0.73817  1st Qu.: -1.214482
## Median : -0.03223  Median : -0.01123  Median : -0.16177  Median : -0.347820
## Mean   : -0.01577  Mean   : -0.01785  Mean   :  0.07852  Mean   : -0.004279
## 3rd Qu.:  0.83810  3rd Qu.:  0.77217  3rd Qu.:  0.66958  3rd Qu.:  0.518842
## Max.    :  1.88251  Max.    :  1.99080  Max.    :  2.91977  Max.    :  1.385504
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.   :-1.10078  Min.   :-0.55446  Min.   :-0.34003  Min.   :-0.253349
## 1st Qu.: -0.69515  1st Qu.: -0.55446  1st Qu.: -0.34003  1st Qu.: -0.253349
## Median : -0.17361  Median : -0.55446  Median : -0.34003  Median : -0.253349
## Mean   :  0.07535  Mean   :  0.04176  Mean   : -0.02811  Mean   : -0.001148
## 3rd Qu.:  0.40587  3rd Qu.:  0.53920  3rd Qu.: -0.34003  3rd Qu.: -0.253349
## Max.    :  4.69404  Max.    :  5.57998  Max.    :  2.93997  Max.    :  3.945807
##      Online      CreditCard      Education_1      Education_2
## Min.   :-1.22199  Min.   :-0.65610  Min.   :-0.82489  Min.   :-0.63693
## 1st Qu.: -1.22199  1st Qu.: -0.65610  1st Qu.: -0.82489  1st Qu.: -0.63693
## Median :  0.81806  Median : -0.65610  Median : -0.82489  Median : -0.63693
## Mean   : -0.06004  Mean   : -0.04516  Mean   :  0.08238  Mean   : -0.05605
## 3rd Qu.:  0.81806  3rd Qu.:  1.52364  3rd Qu.:  1.21188  3rd Qu.:  1.56952
## Max.    :  0.81806  Max.    :  1.52364  Max.    :  1.21188  Max.    :  1.56952
##      Education_3      Personal_Loan
## Min.   :-0.66443      0:902
## 1st Qu.: -0.66443      1: 97
## Median : -0.66443
## Mean   : -0.03263
## 3rd Qu.:  1.50455
## Max.    :  1.50455
```

```
var(valid.norm.df[, 1:13])
```

```

##           Age      Experience      Income      Family
## Age      0.988640890  0.983620705 -0.05324144 -0.0291677024
## Experience 0.983620705  0.991389432 -0.04230932 -0.0416402612
## Income    -0.053241441 -0.042309325  1.05976315 -0.1719799311
## Family    -0.029167702 -0.041640261 -0.17197993  0.9716375312
## CCAvg     -0.089843407 -0.089038093  0.71890864 -0.1310234800
## Mortgage  0.013309492  0.015918091  0.23962745 -0.0084001235
## Securities.Account 0.003440234  0.005214413 -0.02263717  0.0009751088
## CD.Account -0.006720266 -0.005653175  0.12990686  0.0664262003
## Online    -0.012213226 -0.016674470 -0.01292899  0.0752044456
## CreditCard -0.013491343 -0.017623450  0.01409637  0.0378748178
## Education_1 0.007248919  0.037990081  0.23978179 -0.1298181906
## Education_2 -0.034973547 -0.035985475 -0.14510972  0.1336417328
## Education_3 0.026660240 -0.005081626 -0.11270059  0.0068723665
##           CCAvg      Mortgage Securities.Account  CD.Account
## Age      -0.089843407  0.013309492      0.0034402338 -0.006720266
## Experience -0.089038093  0.015918091      0.0052144127 -0.005653175
## Income    0.718908643  0.239627446      -0.0226371651  0.129906860
## Family    -0.131023480 -0.008400124      0.0009751088  0.066426200
## CCAvg     1.142585679  0.126223438      -0.0173574250  0.117260515
## Mortgage  0.126223438  1.108702229      0.0184666092  0.136166383
## Securities.Account -0.017357425  0.018466609      0.9267071962  0.321480264
## CD.Account 0.117260515  0.136166383      0.3214802639  0.996425614
## Online    -0.007905135  0.005950674      0.0059732302  0.187345897
## CreditCard -0.040087709  0.023675047      -0.0044890799  0.295165953
## Education_1 0.158685238  0.031931752      0.0246517999  0.002341910
## Education_2 -0.108116750 -0.037682693      -0.0435823481 -0.025955515
## Education_3 -0.062704782  0.003038356      0.0165903389  0.023020864
##           Online CreditCard Education_1 Education_2
## Age      -0.012213226 -0.01349134  0.007248919 -0.034973547
## Experience -0.016674470 -0.01762345  0.037990081 -0.035985475
## Income    -0.012928991  0.01409637  0.239781791 -0.145109725
## Family    0.075204446  0.03787482 -0.129818191  0.133641733
## CCAvg     -0.007905135 -0.04008771  0.158685238 -0.108116750
## Mortgage  0.005950674  0.02367505  0.031931752 -0.037682693
## Securities.Account 0.005973230 -0.00448908  0.024651800 -0.043582348
## CD.Account 0.187345897  0.29516595  0.002341910 -0.025955515
## Online    1.021336161  0.03350928 -0.026889452 -0.003593782
## CreditCard 0.033509283  0.95940600 -0.025466628  0.011017893
## Education_1 -0.026889452 -0.02546663  1.025786474 -0.527538363
## Education_2 -0.003593782  0.01101789 -0.527538363  0.945198332
## Education_3 0.032167642  0.01628889 -0.573790886 -0.367367285
##           Education_3
## Age      0.026660240
## Experience -0.005081626
## Income    -0.112700587
## Family    0.006872366
## CCAvg     -0.062704782
## Mortgage  0.003038356
## Securities.Account 0.016590339
## CD.Account 0.023020864
## Online    0.032167642
## CreditCard 0.016288892

```

```
## Education_1      -0.573790886
## Education_2      -0.367367285
## Education_3       0.972165816
```

```
dim(train.norm.df[, 1:13]) # Train is 13 numeric variable
```

```
## [1] 3000  13
```

```
dim(test.norm.df)      # Test is 13 numeric variable
```

```
## [1] 1001  13
```

```
table(train.norm.df[, 14]) # DV is logic (Y/N) variable
```

```
##
##    0    1
## 2718 282
```

```
library(FNN)
nn <- knn(train = train.norm.df[, 1:13], test = test.norm.df,
          cl = train.norm.df[, 14], k = 1, prob=TRUE) # We use k = #Q1

customer_df<- data.frame("Age" =40, "Experience"=10, "Income"=84, "Family"=2, "CCAvg"=2, "Mortga
ge"=0, "Securities Account"=0, "CD Account"=0, "Online" =1, "Credit Card"=1, "Education_1" = 0
, "Education_2" = 1, "Education_3" = 0)

customerClass <- knn(train.norm.df[, 1:13], customer_df, cl = train.norm.df[, 14], k = 1, prob
= 0.5)

paste0("The customer is classified as: ", summary(customerClass))
```

```
## [1] "The customer is classified as: 1"
```

```
# print(nn) # uncomment for more output
# row.names(Train_Data)[attr(nn, "nn.index")]

#table(nn)
#table(train.norm.df[, 14])
#table(valid.norm.df[, 14])

#nn <- knn(train.norm.df[, 1:13], valid.norm.df[, 1:13], cl = train.norm.df[, 14], k = 1)

#confusionMatrix(nn, valid.norm.df[, 14])$overall[1]
```

Q1 Answer: The customer is classified as: 1 (loan acceptance)

```
#Q2
# initialize a data frame with two columns: k, and accuracy.
library(caret)
accuracy.df <- data.frame(k = seq(1, 14, 1), accuracy = rep(0, 14))
# compute knn for different k on validation.
for(i in 1:14) {
  knn.pred <- knn(train.norm.df[, 1:13], valid.norm.df[, 1:13],
                  cl = train.norm.df[, 14], k = i)
  accuracy.df[i, 2] <- confusionMatrix(knn.pred, valid.norm.df[, 14])$overall[1]
}
accuracy.df
```

```
##      k  accuracy
## 1    1 0.9489489
## 2    2 0.9499499
## 3    3 0.9549550
## 4    4 0.9529530
## 5    5 0.9529530
## 6    6 0.9519520
## 7    7 0.9539540
## 8    8 0.9499499
## 9    9 0.9529530
## 10  10 0.9439439
## 11  11 0.9469469
## 12  12 0.9439439
## 13  13 0.9439439
## 14  14 0.9419419
```

The value of k that provides the best performance is k = 3. k=3 gives the highest accuracy percentage of 95.5% however it should be noted that this classification model is built based on actual converted customers percentage of only 9.8% of the whole dataset which might be causing overfitting in the model

*#Q3 - Confusion Matrix*

```
knn.pred <- knn(train.norm.df[, 1:13], valid.norm.df[, 1:13],
                 cl = train.norm.df[, 14], k = 3)
confusionMatrix(knn.pred, valid.norm.df[, 14])$overall[1]
```

```
## Accuracy
## 0.954955
```

```
confusionMatrix(knn.pred, valid.norm.df[, 14])
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 897  40
##           1   5  57
##
##           Accuracy : 0.955
##           95% CI : (0.9402, 0.967)
##    No Information Rate : 0.9029
##    P-Value [Acc > NIR] : 6.550e-10
##
##           Kappa : 0.6938
##
##    McNemar's Test P-Value : 4.011e-07
##
##           Sensitivity : 0.9945
##           Specificity : 0.5876
##           Pos Pred Value : 0.9573
##           Neg Pred Value : 0.9194
##           Prevalence : 0.9029
##           Detection Rate : 0.8979
##    Detection Prevalence : 0.9379
##           Balanced Accuracy : 0.7910
##
##           'Positive' Class : 0
##
```

```
#Q4
customerClass <- knn(train.norm.df[, 1:13], customer_df, cl = train.norm.df[, 14], k = 3, prob
= 0.5)

paste0("The customer is classified as: ", summary(customerClass))
```

```
## [1] "The customer is classified as: 1"
```

The customer is classified as: 1 (loan acceptance)

```
#Q5
set.seed(15)
Test_Index = createDataPartition(uk1$Age, p=0.2, list=FALSE) # 20% reserved for Test
Test_Data = uk1[Test_Index,]
TraVal_Data = uk1[-Test_Index,] # Validation and Training data is rest
Train_Index = createDataPartition(TraVal_Data$Age, p=0.625, list=FALSE) # 62.5% of remaining dat
a as training

# 62.5% of the 80% is 50% which is training

Train_Data = TraVal_Data[Train_Index,]
Validation_Data = TraVal_Data[-Train_Index,] # rest as validation
```

## ## Normalization

```
# Copy the original data
train.norm.df <- Train_Data
valid.norm.df <- Validation_Data
traval.norm.df <- TraVal_Data
# use preProcess() from the caret package to normalize Sales and Age.
norm.values <- preProcess(Train_Data[, 1:13], method=c("center", "scale"))
train.norm.df[, 1:13] <- predict(norm.values, Train_Data[, 1:13]) # Replace first two columns with normalized values
valid.norm.df[, 1:13] <- predict(norm.values, Validation_Data[, 1:13])
traval.norm.df[, 1:13] <- predict(norm.values, traval.norm.df[, 1:13])
test.norm.df <- predict(norm.values, Test_Data[, 1:13])
summary(train.norm.df)
```

```
##      Age      Experience      Income      Family
## Min.   :-1.95829   Min.   :-2.024044   Min.   :-1.4303   Min.   :-1.2228
## 1st Qu.: -0.90687   1st Qu.: -0.884555   1st Qu.: -0.7662   1st Qu.: -1.2228
## Median :-0.03069   Median :-0.008026   Median :-0.2128   Median :-0.3559
## Mean    : 0.00000   Mean    : 0.000000   Mean    : 0.0000   Mean    : 0.0000
## 3rd Qu.: 0.84549   3rd Qu.: 0.868504   3rd Qu.: 0.4955   3rd Qu.: 1.3777
## Max.     : 1.89691   Max.     : 2.007992   Max.     : 3.2181   Max.     : 1.3777
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.   :-1.1015   Min.   :-0.5593   Min.   :-0.3413   Min.   :-0.2552
## 1st Qu.: -0.7026   1st Qu.: -0.5593   1st Qu.: -0.3413   1st Qu.: -0.2552
## Median :-0.2466   Median :-0.5593   Median :-0.3413   Median :-0.2552
## Mean    : 0.0000   Mean    : 0.0000   Mean    : 0.0000   Mean    : 0.0000
## 3rd Qu.: 0.3803   3rd Qu.: 0.4566   3rd Qu.: -0.3413   3rd Qu.: -0.2552
## Max.     : 4.1989   Max.     : 5.8919   Max.     : 2.9290   Max.     : 3.9167
##      Online      CreditCard      Education_1      Education_2
## Min.   :-1.1860   Min.   :-0.6531   Min.   :-0.8256   Min.   :-0.6307
## 1st Qu.: -1.1860   1st Qu.: -0.6531   1st Qu.: -0.8256   1st Qu.: -0.6307
## Median : 0.8428   Median :-0.6531   Median :-0.8256   Median :-0.6307
## Mean    : 0.0000   Mean    : 0.0000   Mean    : 0.0000   Mean    : 0.0000
## 3rd Qu.: 0.8428   3rd Qu.: 1.5306   3rd Qu.: 1.2107   3rd Qu.: 1.5848
## Max.     : 0.8428   Max.     : 1.5306   Max.     : 1.2107   Max.     : 1.5848
##      Education_3      Personal_Loan
## Min.   :-0.670      0:2264
## 1st Qu.: -0.670      1: 237
## Median :-0.670
## Mean    : 0.000
## 3rd Qu.: 1.492
## Max.     : 1.492
```

```
var(train.norm.df[, 1:13])
```

```

##           Age      Experience      Income      Family
## Age      1.0000000000  0.9940225594 -0.05087467 -0.0448921532
## Experience 0.9940225594  1.0000000000 -0.04161191 -0.0511499669
## Income    -0.0508746670 -0.0416119125  1.00000000 -0.1540511082
## Family    -0.0448921532 -0.0511499669 -0.15405111  1.0000000000
## CCAvg     -0.0499342272 -0.0477227657  0.65470526 -0.1094696043
## Mortgage  -0.0061419578 -0.0030797005  0.17857854 -0.0291903291
## Securities.Account 0.0110696326  0.0094137497 -0.00604324  0.0349508776
## CD.Account  -0.0095900712 -0.0080462575  0.14013474 -0.0026431396
## Online     -0.0007167343 -0.0009863074  0.03006638 -0.0213492131
## CreditCard  0.0029851469  0.0056280970 -0.02879987  0.0029107587
## Education_1 -0.0142139861  0.0107205722  0.20863503 -0.1012372986
## Education_2 -0.0227033797 -0.0228528694 -0.10242607  0.1111009110
## Education_3  0.0372456641  0.0109184348 -0.12155815 -0.0009315549
##           CCAvg      Mortgage Securities.Account      CD.Account
## Age      -0.04993423 -0.006141958      0.0110696326 -0.009590071
## Experience -0.04772277 -0.003079701      0.0094137497 -0.008046258
## Income     0.65470526  0.178578543      -0.0060432404  0.140134735
## Family    -0.10946960 -0.029190329      0.0349508776 -0.002643140
## CCAvg      1.00000000  0.083809670      0.0138911748  0.121924323
## Mortgage  0.08380967  1.000000000      -0.0258171796  0.051877435
## Securities.Account 0.01389117 -0.025817180      1.0000000000  0.365817547
## CD.Account  0.12192432  0.051877435      0.3658175470  1.000000000
## Online     0.01061331  0.005725267      0.0117521686  0.177952540
## CreditCard -0.02317236 -0.026694132      0.0055415949  0.270532380
## Education_1  0.16482286  0.055649900      0.0004814151 -0.010303193
## Education_2 -0.08409744 -0.015173448      0.0020197981  0.012729568
## Education_3 -0.09292824 -0.044276991      -0.0024820985 -0.001482952
##           Online      CreditCard      Education_1      Education_2
## Age      -0.0007167343  0.002985147 -0.0142139861 -0.022703380
## Experience -0.0009863074  0.005628097  0.0107205722 -0.022852869
## Income     0.0300663793 -0.028799867  0.2086350284 -0.102426065
## Family    -0.0213492131  0.002910759 -0.1012372986  0.111100911
## CCAvg      0.0106133117 -0.023172363  0.1648228643 -0.084097439
## Mortgage  0.0057252667 -0.026694132  0.0556499000 -0.015173448
## Securities.Account 0.0117521686  0.005541595  0.0004814151  0.002019798
## CD.Account  0.1779525401  0.270532380 -0.0103031929  0.012729568
## Online     1.0000000000  0.006635667  0.0136334716  0.021196286
## CreditCard  0.0066356667  1.000000000  0.0173110016 -0.023116442
## Education_1  0.0136334716  0.017311002  1.0000000000 -0.520952732
## Education_2  0.0211962862 -0.023116442 -0.5209527318  1.000000000
## Education_3 -0.0351586636  0.004178555 -0.5533423992 -0.422732154
##           Education_3
## Age      0.0372456641
## Experience 0.0109184348
## Income    -0.1215581517
## Family    -0.0009315549
## CCAvg     -0.0929282409
## Mortgage  -0.0442769911
## Securities.Account -0.0024820985
## CD.Account -0.0014829520
## Online    -0.0351586636
## CreditCard 0.0041785553

```

```
## Education_1      -0.5533423992
## Education_2      -0.4227321536
## Education_3       1.0000000000
```

```
summary(valid.norm.df)
```

```
##      Age      Experience      Income      Family
## Min.   :-1.958287  Min.   :-1.936391  Min.   :-1.43027  Min.   :-1.22276
## 1st Qu.: -0.906870  1st Qu.: -0.884555  1st Qu.: -0.72195  1st Qu.: -1.22276
## Median : -0.030689  Median : -0.008026  Median : -0.19071  Median : -0.35594
## Mean   : -0.005889  Mean   : -0.003228  Mean   :  0.03345  Mean   : -0.02438
## 3rd Qu.:  0.845492  3rd Qu.:  0.868504  3rd Qu.:  0.56188  3rd Qu.:  0.51087
## Max.   :  1.896908  Max.   :  2.007992  Max.   :  2.90820  Max.   :  1.37768
##      CCAvg      Mortgage      Securities.Account      CD.Account
## Min.   :-1.101522  Min.   : -0.55930  Min.   : -0.34128  Min.   : -0.25522
## 1st Qu.: -0.702567  1st Qu.: -0.55930  1st Qu.: -0.34128  1st Qu.: -0.25522
## Median : -0.246618  Median : -0.55930  Median : -0.34128  Median : -0.25522
## Mean   : -0.000963  Mean   :  0.04784  Mean   : -0.02473  Mean   : -0.01014
## 3rd Qu.:  0.323319  3rd Qu.:  0.47442  3rd Qu.: -0.34128  3rd Qu.: -0.25522
## Max.   :  4.597842  Max.   :  5.42457  Max.   :  2.92898  Max.   :  3.91667
##      Online      CreditCard      Education_1      Education_2
## Min.   :-1.18598  Min.   : -0.65309  Min.   : -0.82561  Min.   : -0.63074
## 1st Qu.: -1.18598  1st Qu.: -0.65309  1st Qu.: -0.82561  1st Qu.: -0.63074
## Median :  0.84284  Median : -0.65309  Median : -0.82561  Median : -0.63074
## Mean   :  0.03836  Mean   : -0.01898  Mean   :  0.05255  Mean   : -0.01399
## 3rd Qu.:  0.84284  3rd Qu.:  1.53057  3rd Qu.:  1.21074  3rd Qu.:  1.58481
## Max.   :  0.84284  Max.   :  1.53057  Max.   :  1.21074  Max.   :  1.58481
##      Education_3      Personal_Loan
## Min.   : -0.66995  0:1356
## 1st Qu.: -0.66995  1: 142
## Median : -0.66995
## Mean   : -0.04213
## 3rd Qu.:  1.49205
## Max.   :  1.49205
```

```
var(valid.norm.df[, 1:13])
```



```

##           Age      Experience      Income      Family
## Age      1.028410613  1.024109453 -0.04727932 -0.087829557
## Experience 1.024109453  1.031783061 -0.03884993 -0.094776283
## Income    -0.047279315 -0.038849927  1.03384575 -0.159771772
## Family    -0.087829557 -0.094776283 -0.15977177  0.981654164
## CCAvg     -0.074329772 -0.073375939  0.66390120 -0.106480198
## Mortgage  -0.009178939 -0.009777290  0.30546794 -0.027895739
## Securities.Account -0.026230479 -0.023157273  0.01898459 -0.006559310
## CD.Account  0.018091560  0.019830856  0.19461122 -0.006427784
## Online     0.042407374  0.041995080 -0.01318926  0.013170172
## CreditCard  0.043310936  0.041961680  0.01032662  0.007091008
## Education_1 -0.022269432  0.005441586  0.22775413 -0.127464047
## Education_2 -0.013619599 -0.016582389 -0.16956079  0.150729617
## Education_3  0.036933939  0.010404233 -0.07634492 -0.011757385
##           CCAvg      Mortgage Securities.Account  CD.Account
## Age      -0.074329772 -0.009178939      -0.026230479  0.018091560
## Experience -0.073375939 -0.009777290      -0.023157273  0.019830856
## Income     0.663901201  0.305467938      0.018984593  0.194611222
## Family     -0.106480198 -0.027895739      -0.006559310 -0.006427784
## CCAvg      1.007459332  0.189048851      0.008481167  0.130988547
## Mortgage   0.189048851  1.192971189      0.000547136  0.161211890
## Securities.Account 0.008481167  0.000547136      0.935615396  0.268688738
## CD.Account   0.130988547  0.161211890      0.268688738  0.963013440
## Online     -0.019822694 -0.012446249      0.028793633  0.169023601
## CreditCard  0.015692529  0.027311519      -0.033898442  0.288731606
## Education_1  0.154784243  0.045472017      0.051024068 -0.022411918
## Education_2 -0.110284479 -0.077018583      -0.025960708  0.003107789
## Education_3 -0.056716117  0.026879206      -0.028839203  0.020762138
##           Online  CreditCard Education_1 Education_2
## Age      0.042407374  0.043310936 -0.022269432 -0.013619599
## Experience 0.041995080  0.041961680  0.005441586 -0.016582389
## Income    -0.013189264  0.010326618  0.227754126 -0.169560790
## Family     0.013170172  0.007091008 -0.127464047  0.150729617
## CCAvg     -0.019822694  0.015692529  0.154784243 -0.110284479
## Mortgage  -0.012446249  0.027311519  0.045472017 -0.077018583
## Securities.Account 0.028793633 -0.033898442  0.051024068 -0.025960708
## CD.Account  0.169023601  0.288731606 -0.022411918  0.003107789
## Online     0.985625752  0.001450083  0.003194576 -0.001944309
## CreditCard  0.001450083  0.983238995 -0.001752896  0.009399846
## Education_1  0.003194576 -0.001752896  1.017755073 -0.541962242
## Education_2 -0.001944309  0.009399846 -0.541962242  0.986714480
## Education_3 -0.001494379 -0.007311591 -0.551691322 -0.387461883
##           Education_3
## Age      0.036933939
## Experience 0.010404233
## Income    -0.076344922
## Family     -0.011757385
## CCAvg     -0.056716117
## Mortgage   0.026879206
## Securities.Account -0.028839203
## CD.Account  0.020762138
## Online     -0.001494379
## CreditCard -0.007311591

```

```
## Education_1      -0.551691322
## Education_2      -0.387461883
## Education_3       0.963829277
```

## ## KNN

```
library(FNN)
nn <- knn(train = train.norm.df[, 1:13], test = test.norm.df,
          cl = train.norm.df[, 14], k = 3, prob=TRUE)
```

## ## Confusion matrix

```
knn.pred.valid <- knn(train.norm.df[, 1:13], valid.norm.df[, 1:13],
                      cl = train.norm.df[, 14], k = 3)
confusionMatrix(knn.pred.valid, valid.norm.df[, 14])
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1348   56
##           1    8   86
##
##           Accuracy : 0.9573
##           95% CI : (0.9458, 0.9669)
##       No Information Rate : 0.9052
##       P-Value [Acc > NIR] : 1.836e-14
##
##           Kappa : 0.7067
##
##  Mcnemar's Test P-Value : 4.228e-09
##
##           Sensitivity : 0.9941
##           Specificity : 0.6056
##       Pos Pred Value : 0.9601
##       Neg Pred Value : 0.9149
##           Prevalence : 0.9052
##       Detection Rate : 0.8999
##   Detection Prevalence : 0.9372
##       Balanced Accuracy : 0.7999
##
##       'Positive' Class : 0
##
```

```
knn.pred.train <- knn(train.norm.df[, 1:13], train.norm.df[, 1:13],
                      cl = train.norm.df[, 14], k = 3)
confusionMatrix(knn.pred.train, train.norm.df[, 14])
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 2263   58
##           1    1  179
##
##           Accuracy : 0.9764
##           95% CI : (0.9697, 0.982)
##    No Information Rate : 0.9052
##    P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.8459
##
##    McNemar's Test P-Value : 3.086e-13
##
##           Sensitivity : 0.9996
##           Specificity : 0.7553
##           Pos Pred Value : 0.9750
##           Neg Pred Value : 0.9944
##           Prevalence : 0.9052
##           Detection Rate : 0.9048
##    Detection Prevalence : 0.9280
##           Balanced Accuracy : 0.8774
##
##           'Positive' Class : 0
##
```