Assignment 03-Naive Bayes

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```
library(readx1)
 UniversalBank <- read_excel("F:/1st sem/ML/Assignment 03/UniversalBank.xlsx")</pre>
 View(UniversalBank)
install.packages(class) install.packages(e1071) install.packages("reshape") install.packages("reshape2")
install.packages("reshape2") install.packages("gmodels") install.packages("fastDummies")
 library(class)
 library(e1071)
 library(reshape)
 ##
 ## Attaching package: 'reshape'
 ## The following object is masked from 'package:class':
 ##
 ##
         condense
 library(reshape2)
 ##
 ## Attaching package: 'reshape2'
 ## The following objects are masked from 'package:reshape':
 ##
 ##
         colsplit, melt, recast
 library(gmodels)
 library(fastDummies)
 library(caret)
 ## Loading required package: ggplot2
 ## Loading required package: lattice
```

```
library(readr)
 library(dplyr)
 ##
 ## Attaching package: 'dplyr'
 ## The following object is masked from 'package:reshape':
 ##
 ##
        rename
 ## The following objects are masked from 'package:stats':
 ##
        filter, lag
 ##
 ## The following objects are masked from 'package:base':
 ##
 ##
        intersect, setdiff, setequal, union
 library(ISLR)
#Select the required variables
 UniversalBank<-UniversalBank[,c(10,13,14)]</pre>
 t(t(names(UniversalBank)))
 ##
         [,1]
 ## [1,] "Personal Loan"
 ## [2,] "Online"
 ## [3,] "CreditCard"
#Change the column names
 colnames(UniversalBank)[1]="Loan"
 colnames(UniversalBank)[3]="CC"
 t(t(names(UniversalBank)))
 ##
         [,1]
 ## [1,] "Loan"
 ## [2,] "Online"
 ## [3,] "CC"
```

#use 60% of the data in training set & 40% for validation set

```
set.seed(123)
Train_data<-createDataPartition(UniversalBank$Loan,p=.6, list = FALSE)
Train_UB1<-UniversalBank[Train_data,]</pre>
```

```
Valid_UB1=UniversalBank[-Train_data, ]
attach(UniversalBank)
```

A. Create a Pivot table

```
ftable(CC,Loan,Online)
```

```
##
           Online
                      0
                           1
## CC Loan
## 0 0
                  1300 1893
##
                   128 209
      1
## 1
      0
                   527
                         800
##
                          82
      1
                    61
```

B. The probability of customer accepts loan offer

```
P0<-82/882
```

C. Creating two pivot tables: (Loan, Online) & (Loan, CC)

```
addmargins(table(Loan,Online))
```

```
## Online
## Loan 0 1 Sum
## 0 1827 2693 4520
## 1 189 291 480
## Sum 2016 2984 5000
```

```
addmargins(table(Loan,CC))
```

```
## CC
## Loan 0 1 Sum
## 0 3193 1327 4520
## 1 337 143 480
## Sum 3530 1470 5000
```

D. Calculate different probabilities of A given B

```
##1. p(CC=1| Loan = 1)
P1 <- 143/480

##2. P(Online = 1 | Loan = 1)
P2 <- 291/480

##3.P(Loan = 1)
addmargins(table(Loan))</pre>
```

```
## Loan
## 0 1 Sum
## 4520 480 5000
```

```
p3 <- 480/5000

##4.P(CC = 1 | Loan = 0)

p4 <-1327/4520

##5. P(Online = 1 | Loan = 0)

p5<-2693/4520

##6.P(Loan = 0)

p6<-4520/5000
```

E. Naive Bayes of y P(Loan = 1 | CC = 1, Online = 1).

```
P7 = ((P1*P2)*p3/((P1*P2)*p3+(p4*p5)*p6))
P7
```

```
## [1] 0.09881706
```

F. Compare this value with the one obtained from the pivot table in (b)

#Answer: The value for P7 = 0.098 & P0 = 0.092 Due to decimals those two values are sligh tly different, the two values for E. and B. are almost same. Naive Bayes method does not n eeded the exact same independent variable as do by the other method of B. So this cause for this slight difference of values.

G. Run naive Bayes on the data & Compare this to the number you obtained in (e).

```
P8 <- naiveBayes(Loan~.,data = Train_UB1)
P8
```

```
##
## Naive Bayes Classifier for Discrete Predictors
## Call:
## naiveBayes.default(x = X, y = Y, laplace = laplace)
##
## A-priori probabilities:
## Y
##
## 0.90733333 0.09266667
##
## Conditional probabilities:
##
      Online
## Y
                       [,2]
            [,1]
     0 0.5951506 0.4909531
##
     1 0.6438849 0.4797134
##
##
##
      CC
## Y
            [,1]
                       [,2]
##
     0 0.2909625 0.4542897
##
     1 0.3273381 0.4700881
```

Answer :Here, the Naive Bayes provides the same answer as in question (E) .09