

MichaelBasta_Assignment_3

Michael Basta

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```
banks <- read.csv("/Users/michaelbasta/Documents/Fundamentals of Machine Learning /Module 5/UniversalBank.csv")
```

```
#install.packages("reshape")  
#install.packages("reshape2")
```

```
library(reshape)  
library(reshape2)
```

```
##  
## Attaching package: 'reshape2'
```

```
## The following objects are masked from 'package:reshape':  
##  
##   colsplit, melt, recast
```

```
# A.  
table(banks$CreditCard, banks$Personal.Loan, banks$Online)
```

```
## , , = 0  
##  
##  
##      0      1  
## 0 1300 128  
## 1  527   61  
##  
## , , = 1  
##  
##  
##      0      1  
## 0 1893 209  
## 1  800   82
```

```
prop.table(table(banks$CreditCard, banks$Personal.Loan, banks$Online))
```

```
## , , = 0  
##  
##  
##      0      1  
## 0 0.2600 0.0256
```

```
##    1 0.1054 0.0122
##
## , , = 1
##
##
##          0      1
##    0 0.3786 0.0418
##    1 0.1600 0.0164
```

```
# B.
```

```
print("The pobability that the customer will accept the loan conditional on customer has Credit Card and")
```

```
## [1] "The pobability that the customer will accept the loan conditional on customer has Credit Card and"
```

```
# C.
```

```
print("Table row: loan Col: Online")
```

```
## [1] "Table row: loan Col: Online"
```

```
table(banks$Personal.Loan, banks$Online)
```

```
##
##          0      1
##    0 1827 2693
##    1  189  291
```

```
prop.table(table(banks$Personal.Loan, banks$Online))
```

```
##
##          0      1
##    0 0.3654 0.5386
##    1 0.0378 0.0582
```

```
print("Table row: loan Col: Credit Card ")
```

```
## [1] "Table row: loan Col: Credit Card "
```

```
table(banks$Personal.Loan, banks$CreditCard)
```

```
##
##          0      1
##    0 3193 1327
##    1  337  143
```

```
prop.table(table(banks$Personal.Loan, banks$CreditCard))
```

```
##
##          0      1
##    0 0.6386 0.2654
##    1 0.0674 0.0286
```

```

# D.
## i.
print("i.  $P(CC = 1 \mid loan = 1) = 0.286$ ")

## [1] "i.  $P(CC = 1 \mid loan = 1) = 0.286$ "

## ii.
print("ii.  $P(online = 1 \mid loan = 1) = 0.0582$ ")

## [1] "ii.  $P(online = 1 \mid loan = 1) = 0.0582$ "

## iii.
print("iii.  $P(loan = 1) = 0.0378 + 0.0582 = 0.096$ ")

## [1] "iii.  $P(loan = 1) = 0.0378 + 0.0582 = 0.096$ "

## iv.
print("iv.  $P(CC = 1 \mid loan = 0) = 0.2654$ ")

## [1] "iv.  $P(CC = 1 \mid loan = 0) = 0.2654$ "

## v.
print("v.  $P(online = 1 \mid loan = 0) = 0.5386$ ")

## [1] "v.  $P(online = 1 \mid loan = 0) = 0.5386$ "

## vi.
print("vi.  $P(loan = 0) = 0.3654 + 0.5386 = 0.904$ ")

## [1] "vi.  $P(loan = 0) = 0.3654 + 0.5386 = 0.904$ "

# E.
print("Using the long formula 8.3 in textbook")

## [1] "Using the long formula 8.3 in textbook"

print("P(loan = 1 | CC = 1, Online = 1) = (P(loan=1)[P(CC=1 | loan=1) * P(Online=1 | loan=1)] \ (P(loan=1) * P(CC=1 | loan=1))")

## [1] "P(loan = 1 | CC = 1, Online = 1) = (P(loan=1)[P(CC=1 | loan=1) * P(Online=1 | loan=1)] \ (P(loan=1) * P(CC=1 | loan=1))"

sprintf("Numerator = 0.096*[0.286*0.0582] = %1.4f", (0.096*0.286*0.0582))

## [1] "Numerator = 0.096*[0.286*0.0582] = 0.0016"

sprintf("Denominator = 0.002 + 0.904*[0.2654*0.5386] = %1.4f", (0.0016 + (0.904*0.2654*0.5386)))

## [1] "Denominator = 0.002 + 0.904*[0.2654*0.5386] = 0.1308"

```

```
sprintf("Naive Bayes Probability = 0.002/0.131 = %1.4f", (0.002/0.131))
```

```
## [1] "Naive Bayes Probability = 0.002/0.131 = 0.0153"
```

```
# F.
```

```
print("Both values are very close B is 0.016 & E is 0.0153")
```

```
## [1] "Both values are very close B is 0.016 & E is 0.0153"
```

```
#install.packages("caret")
```

```
#install.packages("ISLR")
```

```
#install.packages("e1071")
```

```
library(caret)
```

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

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

```
library(ISLR)
```

```
library(e1071)
```

```
# G.
```

```
# Divide data into test and train (60% - 40%)
```

```
Index_Train <- createDataPartition(banks$Personal.Loan, p=0.6, list=FALSE)
```

```
Train <- banks[Index_Train,]
```

```
Test <- banks[-Index_Train,]
```

```
# Build a naive Bayes classifier
```

```
nb_model <- naiveBayes(Personal.Loan~CreditCard+Online, data = Train)
```

```
# Predict the default status of test dataset
```

```
Predicted_Test_labels <- predict(nb_model, Train, type = "raw")
```

```
head(cbind(Predicted_Test_labels, Train$CreditCard, Train$Online))
```

```
##           0           1
## [1,] 0.9042180 0.09578198 0 0
## [2,] 0.9017867 0.09821328 0 1
## [3,] 0.9017867 0.09821328 0 1
## [4,] 0.9042180 0.09578198 0 0
## [5,] 0.9042180 0.09578198 0 0
## [6,] 0.9042180 0.09578198 0 0
```

```
p <- predict(nb_model, Train)
```

```
t1 <- table(p, Train$CreditCard)
```

```
prop.table(t1)
```

```
##
```

```
## p           0           1
```

```
## 0 0.708 0.292
```

```
## 1 0.000 0.000
```

```
t2 <- table(p,Train$Online)
prop.table(t2)
```

```
##
## p          0          1
## 0 0.3986667 0.6013333
## 1 0.0000000 0.0000000
```

```
#install.packages("gmodels")
library("gmodels")

p <- predict(nb_model,Test)
CrossTable(x=Test$Personal.Loan,y=p,prop.chisq=FALSE)
```

```
##
##
##      Cell Contents
## |-----|
## |                      N |
## |      N / Table Total |
## |-----|
##
##
## Total Observations in Table:  2000
##
##
##      | p
## Test$Personal.Loan |      0 | Row Total |
## -----|-----|-----|
##              0 |      1800 |      1800 |
##              |      0.900 |      |
## -----|-----|-----|
##              1 |       200 |       200 |
##              |      0.100 |      |
## -----|-----|-----|
##      Column Total |      2000 |      2000 |
## -----|-----|-----|
##
##
```