Machine learning Assingment 3

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#Assignment 3 Machine Learning
setwd("C:/Users/nikes/Downloads/Machine Learning Assingment/Assingment 3")
library(readr)
UniversalBank <- read_csv("UniversalBank.csv")</pre>
colnames(UniversalBank) <- c('ID', 'Age', 'Experience', 'Income', 'ZIP_Code', 'Family',
'CCAvg', 'Education', 'Mortgage', 'Personal_Loan', 'Securities_Account', 'CD_Account',
'Online', 'CreditCard')
summary(UniversalBank)
library(caret)
library(dplyr)
library(class)
library(reshape2)
library(ISLR)
library(e1071)
A. Pivot table for Universal Bank
UniversalBankPersonal_Loan = as. factor(UniversalBankPersonal_Loan)
UniversalBankOnline = as. factor(UniversalBankOnline)
UniversalBankCreditCard = as. factor(UniversalBankCreditCard)
set.seed(123)
train.index <- sample(row.names(UniversalBank), 0.6*dim(UniversalBank)[1])
test.index <- setdiff(row.names(UniversalBank), train.index)</pre>
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train.df <- UniversalBank[train.index,]</pre>

test.df <- UniversalBank[test.index,]</pre>

train <- UniversalBank[train.index,]</pre>

 $test = UniversalBank[train.index,] \\ melted.UniversalBank = melt(train, id=c("CreditCard", "Personal_Loan"), variable= "Online") \\ recast.UniversalBank= dcast(melted.UniversalBank, Personal_Loan+CreditCard \sim Online) \\ recast.UniversalBank[,c(1:2,14)]$

B. Probability of the customer accepting loan offer

Probablity = 91 / 3000 = 0.30

C. Separate pivot table

 $table(Personal_Loan=trainPersonal_Loan,Online=trainOnline)$ $table(Personal_Loan=trainPersonal_Loan,CreditCard=trainCreditCard)$

D. [P(A | B) means "the probability of A given B"]

- i. Proportion of credit card holders among the loan acceptors = 91/278 = 0.32
- ii. P(Online = 1 | Loan = 1) = 179/278 = 0.64
- iii. P(Loan = 1) (the proportion of loan acceptors) = 278/2722 = 0.10
- iv. $P(CC = 1 \mid Loan = 0) = 792/2722 = 0.29$
- v. P(Online = 1 | Loan = 0) = 1620/2722 = 0.59
- vi. P(Loan = 0) = 2722/3000 = 0.90

E. naive Bayes probability P(Loan = 1 | CC= 1, Online = 1).

= (0.32*0.64*0.1) / (0.32*0.64*0.1 + 0.29*0.59*0.9)= 0.11

F. Comparing value with the one obtained from the pivot table in (B).

Pivot table Probablity = (278/3000) = 0.092

Naive Bayes Probablity = 0.320.590.1 / (0.320.640.1 + 0.290.590.9) = 0.11

##As using the naive bayes the main assumption we are making is all variable are independent and have equal importance, so we can see naive bayes probablity is little

higher, the accuracy of naive bayes probablity may be less accurate considering the features that all variables are independent and are not corelated with each other.

G. Running naive Bayes on the data. P(Loan = 1 | CC = 1, Online = 1)

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\label{eq:naive.train} $$ naive.train = train.df[,c(10,13:14)]$ $$ naive.test = test.df[,c(10,13:14)]$ $$ naivebayes = naiveBayes(Personal_Loan~.,data=naive.train)$ $$ naivebayes
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We can see and analyze the result the prior probablity is exactly 0.092 as shown here which is exactly the same as we computed above.