

Noah Sprunk Assignment 3

Code ▾

This is an R Markdown (<http://rmarkdown.rstudio.com>) Notebook. When you execute code within the notebook, the results appear beneath the code.

Try executing this chunk by clicking the *Run* button within the chunk or by placing your cursor inside it and pressing *Ctrl+Shift+Enter*.

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```
library(ggplot2)
library(lattice)
library(caret)
library(class)
library(ISLR)
library(e1071)
library(gmodels)

Bank <- read.csv("C:\\\\Users\\\\Noah\\\\Downloads\\\\UniversalBank.csv")

summary(Bank)
```

ID	Age	Experience	Income	ZIP.Code	Family
CCAvg	Education				
Min. : 1	Min. :23.00	Min. :-3.0	Min. : 8.00	Min. : 9307	Min. :1.000
Min. : 0.000	Min. :1.000				
1st Qu.:1251	1st Qu.:35.00	1st Qu.:10.0	1st Qu.: 39.00	1st Qu.:91911	1st Qu.:1.000
1st Qu.: 0.700	1st Qu.:1.000				
Median :2500	Median :45.00	Median :20.0	Median : 64.00	Median :93437	Median :2.000
Median : 1.500	Median :2.000				
Mean :2500	Mean :45.34	Mean :20.1	Mean : 73.77	Mean :93153	Mean :2.396
Mean : 1.938	Mean :1.881				
3rd Qu.:3750	3rd Qu.:55.00	3rd Qu.:30.0	3rd Qu.: 98.00	3rd Qu.:94608	3rd Qu.:3.000
3rd Qu.: 2.500	3rd Qu.:3.000				
Max. :5000	Max. :67.00	Max. :43.0	Max. :224.00	Max. :96651	Max. :4.000
Max. :10.000	Max. :3.000				
Mortgage	Personal.Loan	Securities.Account	CD.Account	Online	Credit
Card					
Min. : 0.0	Min. :0.000	Min. :0.0000	Length:5000	Min. :0.0000	Min. :
0.000					
1st Qu.: 0.0	1st Qu.:0.000	1st Qu.:0.0000	Class :character	1st Qu.:0.0000	1st Qu.:
0.000					
Median : 0.0	Median :0.000	Median :0.0000	Mode :character	Median :1.0000	Median :
0.000					
Mean : 56.5	Mean :0.096	Mean :0.1044		Mean :0.5968	Mean :
0.294					
3rd Qu.:101.0	3rd Qu.:0.000	3rd Qu.:0.0000		3rd Qu.:1.0000	3rd Qu.:
1.000					
Max. :635.0	Max. :1.000	Max. :1.0000		Max. :1.0000	Max. :
1.000					

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```
Bank$Loan_Category='Accepted'  
Bank$Loan_Category[Bank$Personal.Loan < 1]='Declined'  
  
Bank$CreditCard=as.factor(Bank$CreditCard)  
Bank$Online=as.factor(Bank$Online)  
  
Bank$Loan_Category=as.factor(Bank$Loan_Category)  
Bank$Personal.Loan<-NULL  
summary(Bank)
```

	ID	Age	Experience	Income	ZIP.Code	Family
CCAvg		Education				
	Min. : 1	Min. :23.00	Min. :-3.0	Min. : 8.00	Min. : 9307	Min. :1.000
	Min. : 0.000	Min. :1.000				
	1st Qu.:1251	1st Qu.:35.00	1st Qu.:10.0	1st Qu.: 39.00	1st Qu.:91911	1st Qu.:1.000
	1st Qu.: 0.700	1st Qu.:1.000				
	Median :2500	Median :45.00	Median :20.0	Median : 64.00	Median :93437	Median :2.000
	Median : 1.500	Median :2.000				
	Mean :2500	Mean :45.34	Mean :20.1	Mean : 73.77	Mean :93153	Mean :2.396
	Mean : 1.938	Mean :1.881				
	3rd Qu.:3750	3rd Qu.:55.00	3rd Qu.:30.0	3rd Qu.: 98.00	3rd Qu.:94608	3rd Qu.:3.000
	3rd Qu.: 2.500	3rd Qu.:3.000				
	Max. :5000	Max. :67.00	Max. :43.0	Max. :224.00	Max. :96651	Max. :4.000
	Max. :10.000	Max. :3.000				
	Mortgage	Securities.Account	CD.Account	Online	CreditCard	Loan_Category
	Min. : 0.0	Min. :0.0000	Length:5000	0:2016	0:3530	Accepted: 480
	1st Qu.: 0.0	1st Qu.:0.0000	Class :character	1:2984	1:1470	Declined:4520
	Median : 0.0	Median :0.0000	Mode :character			
	Mean : 56.5	Mean :0.1044				
	3rd Qu.:101.0	3rd Qu.:0.0000				
	Max. :635.0	Max. :1.0000				

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```
Train_Index = createDataPartition(Bank$Loan_Category,p=0.6, list=FALSE)  
Train.df=Bank[Train_Index,]  
Validation.df=Bank[-Train_Index,]  
  
summary(Train.df)
```

	ID	Age	Experience	Income	ZIP.Code	Family
CCAvg	Education					
Min.	: 2	Min. :23.00	Min. :-3.00	Min. : 8.00	Min. : 9307	Min. :1.000
Min.	: 0.000	Min. :1.000				
1st Qu.:	1230	1st Qu.:35.00	1st Qu.:10.00	1st Qu.: 39.00	1st Qu.:91789	1st Qu.:1.000
1st Qu.:	0.700	1st Qu.:1.000				
Median :	2482	Median :46.00	Median :21.00	Median : 65.00	Median :93460	Median :2.000
Median :	1.600	Median :2.000				
Mean :	2485	Mean :45.51	Mean :20.27	Mean : 74.43	Mean :93132	Mean :2.391
Mean :	1.958	Mean :1.879				
3rd Qu.:	3736	3rd Qu.:55.00	3rd Qu.:30.00	3rd Qu.: 98.00	3rd Qu.:94609	3rd Qu.:3.000
3rd Qu.:	2.600	3rd Qu.:3.000				
Max. :	5000	Max. :67.00	Max. :43.00	Max. :224.00	Max. :96651	Max. :4.000
Max. :	10.000	Max. :3.000				
Mortgage		Securities.Account	CD.Account	Online	CreditCard	Loan_Category
Min. :	0.00	Min. :0.0000	Length:3000	0:1198	0:2127	Accepted: 288
1st Qu.:	0.00	1st Qu.:0.0000	Class :character	1:1802	1: 873	Declined:2712
Median :	0.00	Median :0.0000	Mode :character			
Mean :	55.98	Mean :0.1043				
3rd Qu.:	98.00	3rd Qu.:0.0000				
Max. :	635.00	Max. :1.0000				

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Task A

```
mytable <- xtabs(~ Online+CreditCard+Loan_Category, data=Train.df)
ftable(mytable)
```

		Loan_Category Accepted Declined	
Online	CreditCard		
0	0	81	770
	1	34	313
1	0	129	1147
	1	44	482

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Task B

```
# If we look at the pivot table from task A,
# we can do the math: 41+515=556
# then, our accepted, online, with CC
# 41/556=0.07374 or 7.374%
```

Task C

```
table(Loan_Category=Train.df$Loan_Category, Online=Train.df$Online)
```

Online

```
Loan_Category 0 1
Accepted    115 173
Declined   1083 1629
```

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```
table(Loan_Category=Train.df$Loan_Category, CreditCard=Train.df$CreditCard)
```

CreditCard

```
Loan_Category 0 1
Accepted    210 78
Declined   1917 795
```

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```
#Task D
```

```
# P(CC = 1 | Loan = 1) (the proportion of credit card holders among the loan acceptors)
```

```
i = 80/(208+80)
```

```
i
```

```
[1] 0.2777778
```

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```
# P(Online = 1 | Loan = 1)
```

```
ii = 168/(120+168)
```

```
ii
```

```
[1] 0.5833333
```

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```
# P(Loan = 1) (the proportion of loan acceptors)
```

```
iii = (120+168)/((1068+1644)+(120+168))
```

```
iii
```

```
[1] 0.096
```

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```
# P(CC = 1 | Loan = 0)
```

```
iv = 822/(822+1890)
```

```
iv
```

```
[1] 0.3030973
```

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```
# P(Online = 1 | Loan = 0)
v = 1644/(1644+1068)
v
```

```
[1] 0.6061947
```

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```
# P(Loan = 0)
vi = (1890+822)/((1890+822)+(208+80))
vi
```

```
[1] 0.904
```

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```
# Task E
E = (i * iii ) / ((80+822) / ((80+822) + (208+1890)))
E
```

```
[1] 0.0886918
```

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```
# Task F
# Which is more accurate, Task B or E?
# I believe that Task B was more accurate because we had to use less
# probabilities in that calculation. It is less accurate because the
# probabilities used may not be entirely independent. We are assuming that
# they are independent but realistically there are correlations
# between these aspects of finances.
```

```
# Task G
nb.model<-naiveBayes (Loan_Category~CreditCard+Online, data=Train.df)
To_Predict=data.frame(CreditCard='1', Online='1')
predict(nb.model, To_Predict, type='raw')
```

```
Accepted Declined
[1,] 0.08935124 0.9106488
```

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```
# The number that I got in E was 8.87% compared to 8.63% in Task G.  
# This shows me that if we used naive Bayes for both but got two  
# different answers. We took the raw numbers of accepted vs. declined in  
# Task E vs. in Task G we took the training data. They are very close together  
# though which tells me that they are both accurate. I think that  
# the Task E number is more accurate though because it used the  
# raw counts instead of a partition trained data set. But it does tell  
# me that the data set is very accurate when it needs to predict an outcome.
```

Add a new chunk by clicking the *Insert Chunk* button on the toolbar or by pressing *Ctrl+Alt+I*.

When you save the notebook, an HTML file containing the code and output will be saved alongside it (click the *Preview* button or press *Ctrl+Shift+K* to preview the HTML file).

The preview shows you a rendered HTML copy of the contents of the editor. Consequently, unlike *Knit*, *Preview* does not run any R code chunks. Instead, the output of the chunk when it was last run in the editor is displayed.