

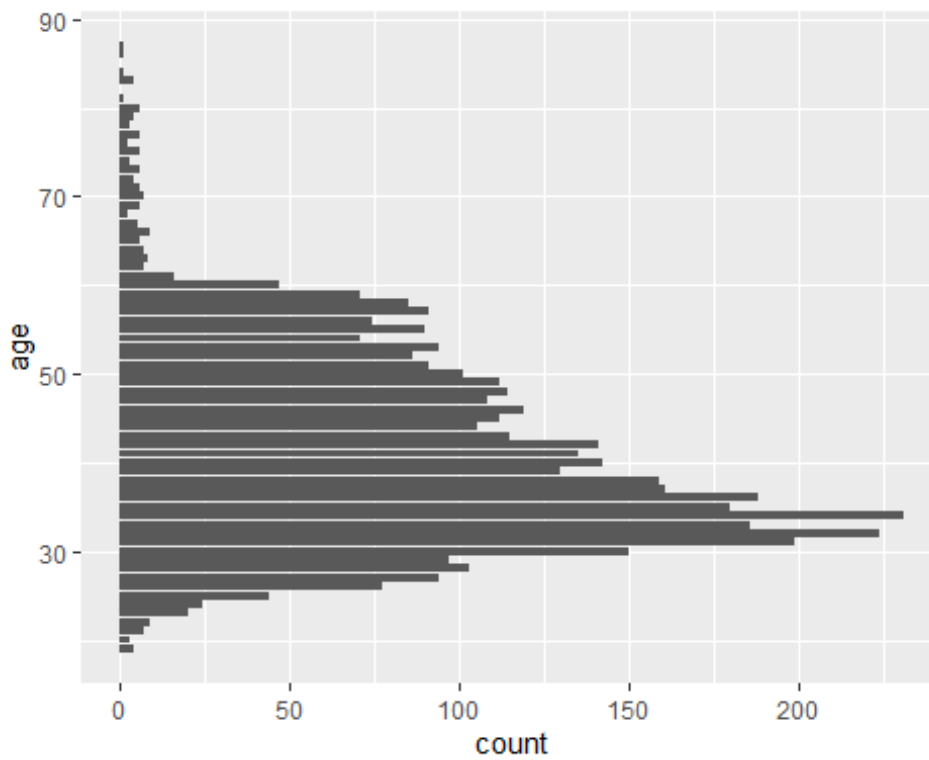
## Part3.R

Cheng Jun

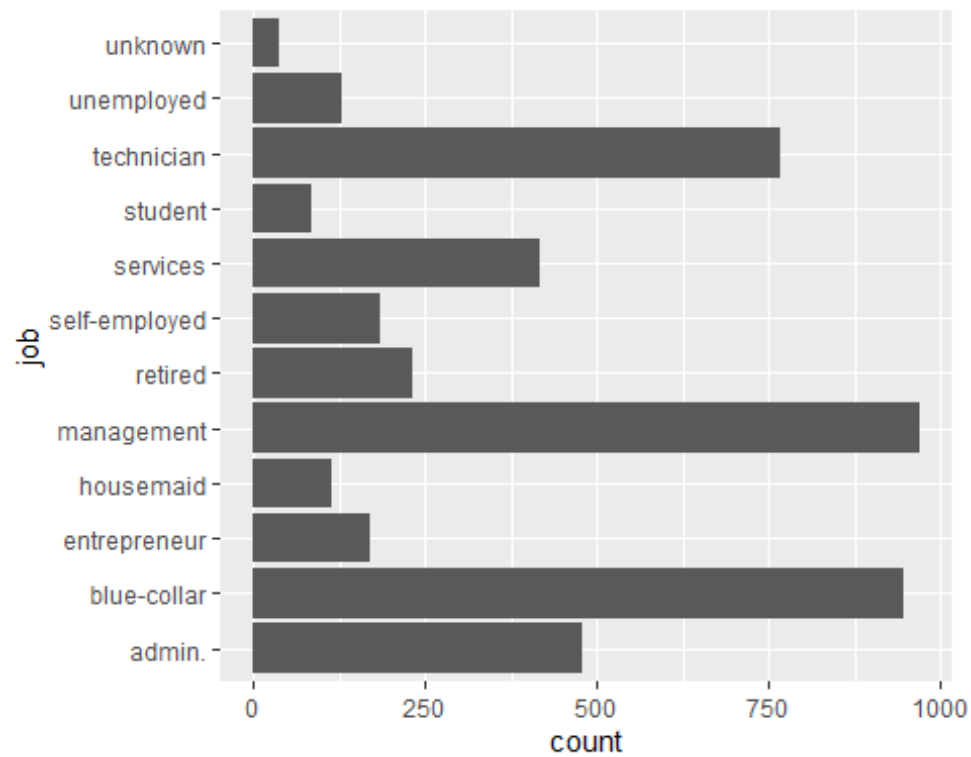
2021-03-26

```
library(caTools)
## Warning: package 'caTools' was built under R version 4.0.4
library(ggplot2)
## Warning: package 'ggplot2' was built under R version 4.0.4
library(randomForest)
## Warning: package 'randomForest' was built under R version 4.0.4
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:ggplot2':
##
##     margin
library(caret)
## Warning: package 'caret' was built under R version 4.0.4
## Loading required package: lattice
## Warning: package 'lattice' was built under R version 4.0.4
setwd("C:/Users/Cheng Jun/Desktop/SIM/Year 2/Machine Learning/Coursework")
data <- read.table("bank.csv",header=1,sep = ';')
sum(is.na(data))
## [1] 0
columns = names(data)
columns
## [1] "age"      "job"      "marital"  "education" "default"  "balance"
## [7] "housing"  "loan"     "contact"  "day"       "month"
## [13] "duration"
## [13] "campaign" "pdays"   "previous" "poutcome" "y"
```

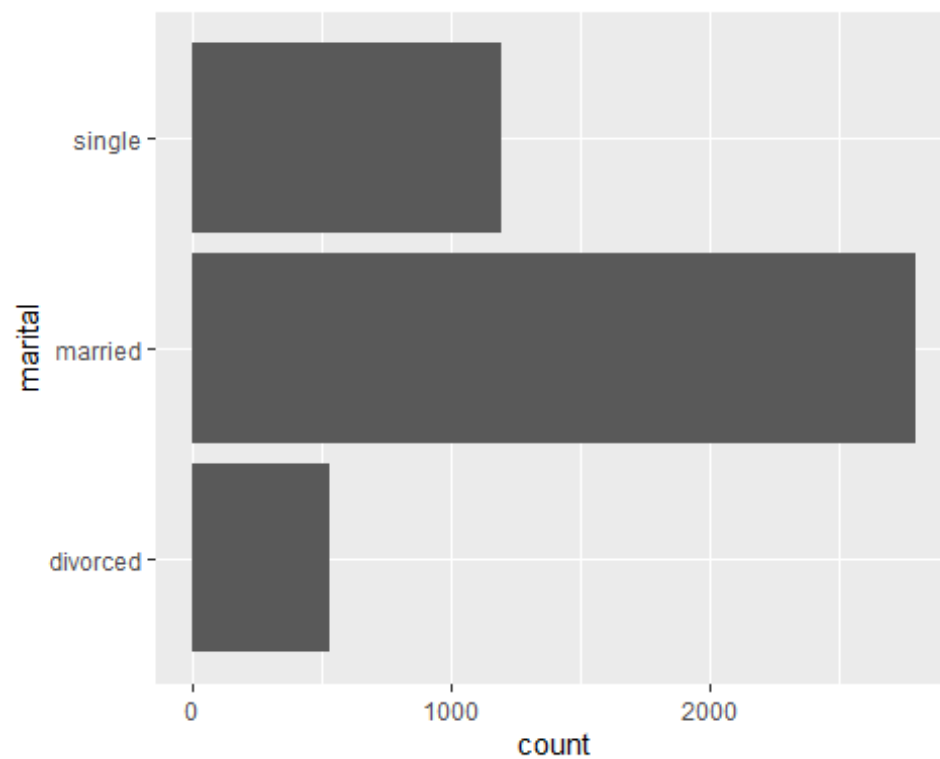
```
#Illustration of data  
#plot age row  
ggplot(data) + geom_bar(aes(y = age))
```



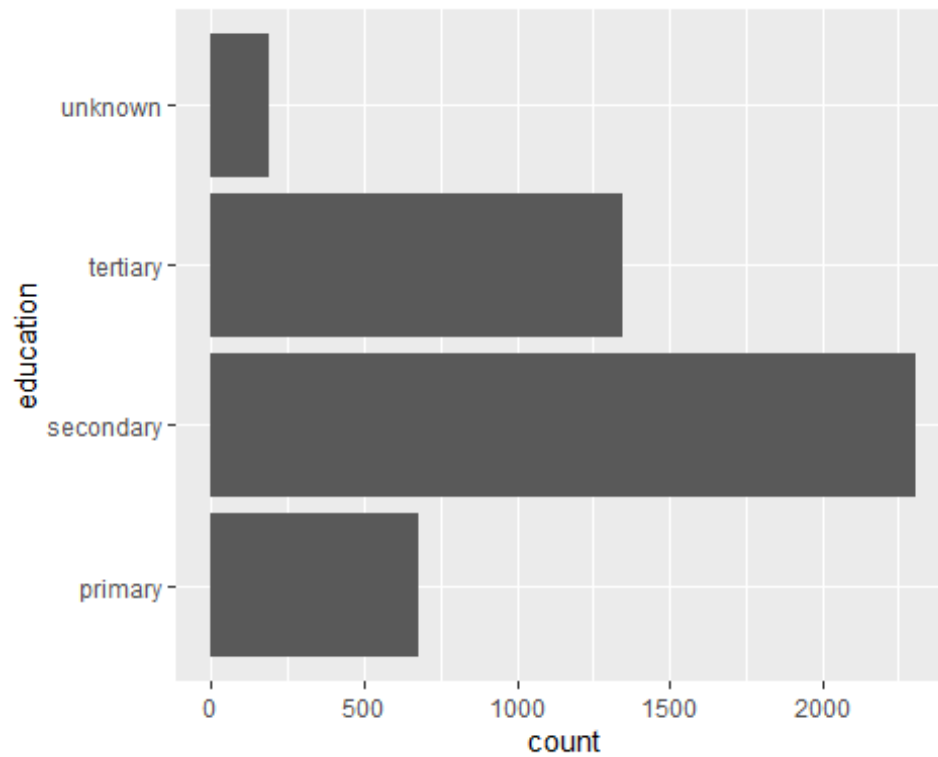
```
#plot job row  
ggplot(data) + geom_bar(aes(y = job))
```



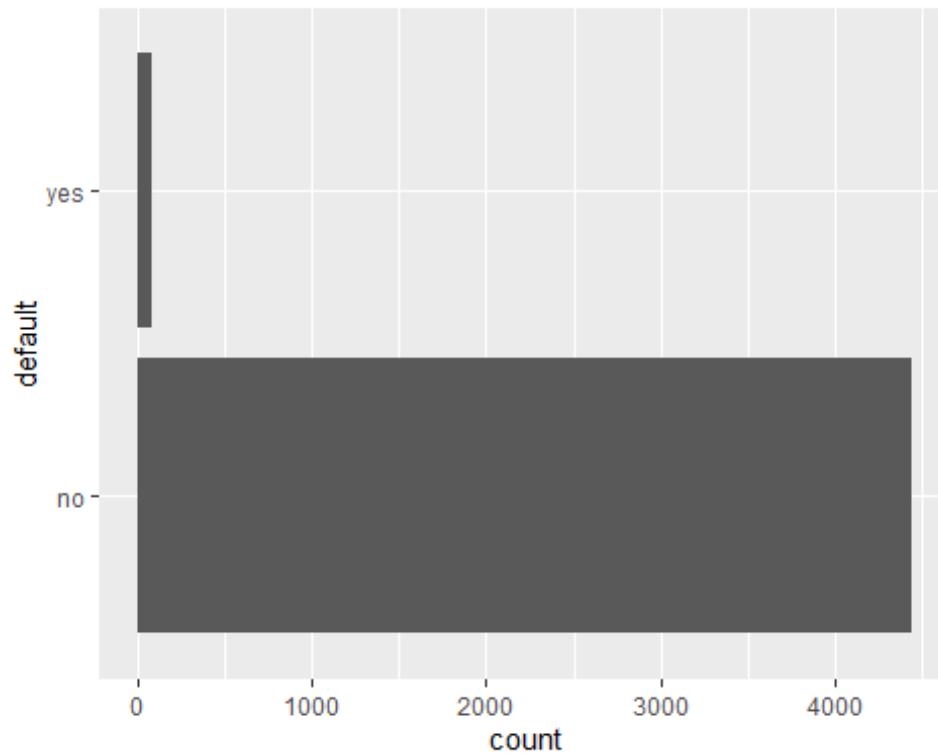
```
#plot marital row  
ggplot(data) + geom_bar(aes(y = marital))
```



```
#plot education row  
ggplot(data) + geom_bar(aes(y = education))
```



```
#plot default row  
ggplot(data) + geom_bar(aes(y = default))
```

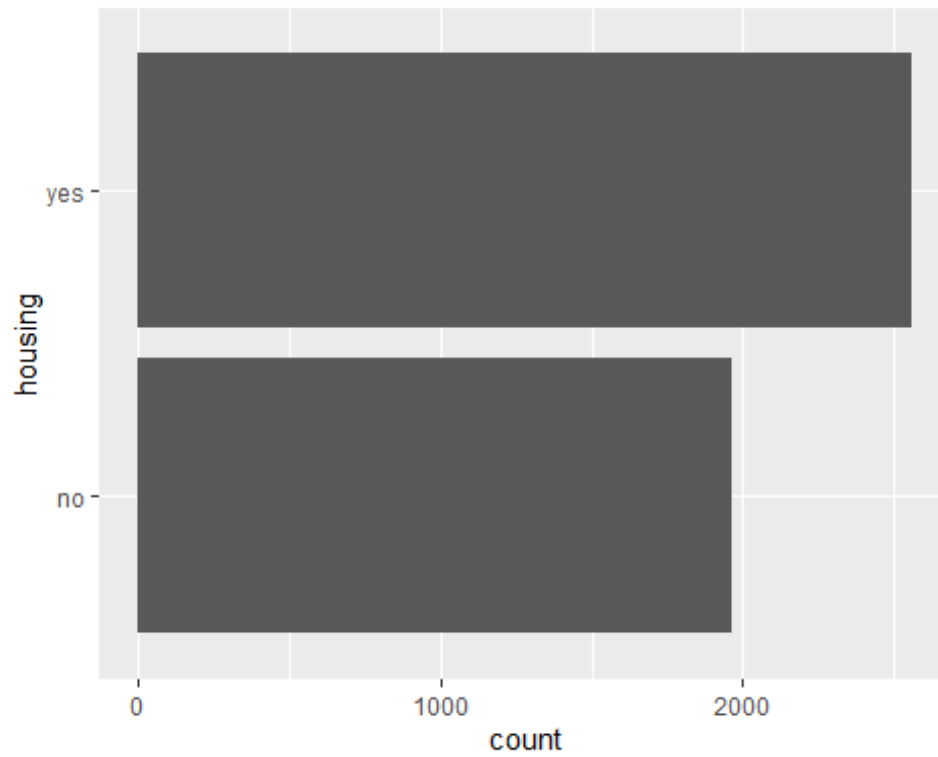


*#As default has mostly 'no' value if we place it in the dataframe it might result in overfitting due to it not being equally distributed.*

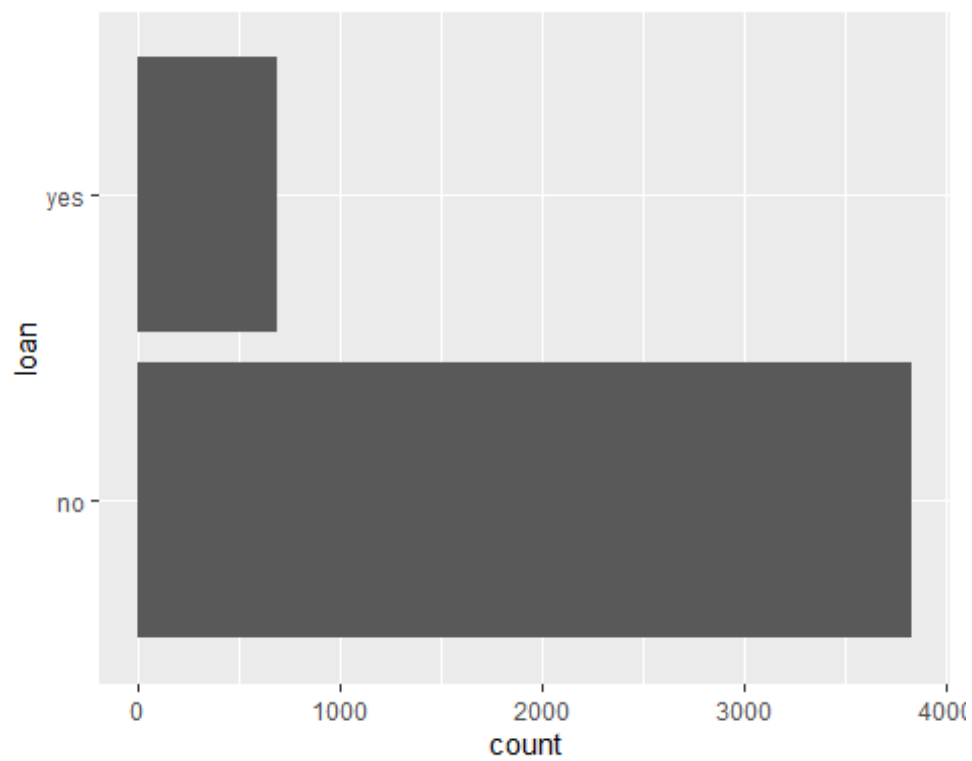
```
data = data[,-c(5)]
```

```
#plot housing row
```

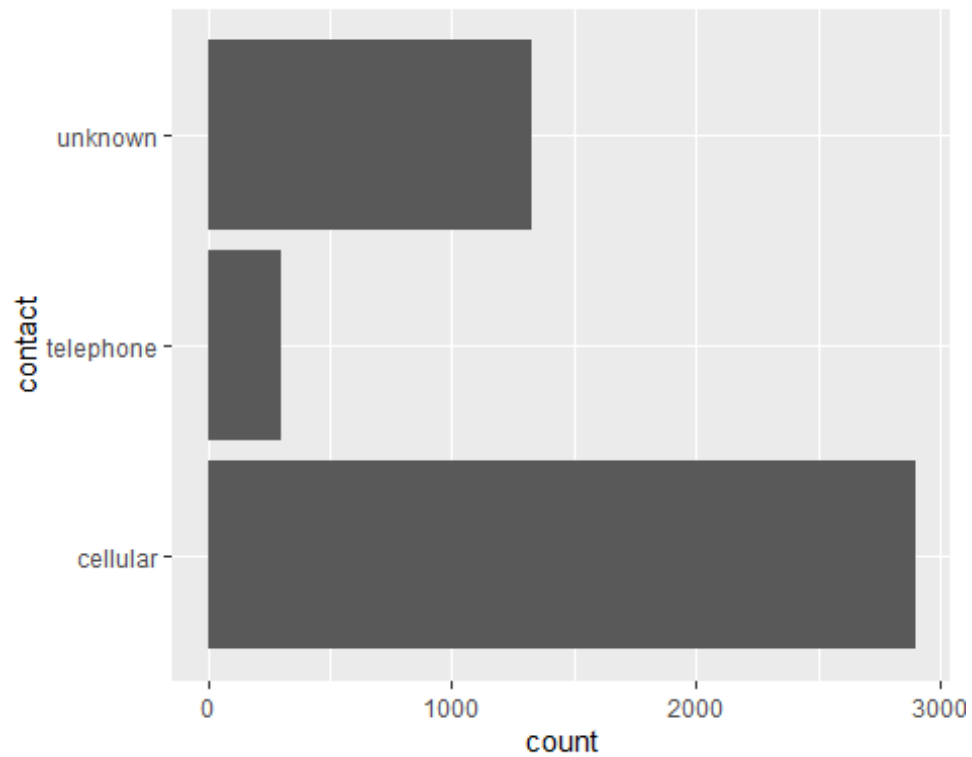
```
ggplot(data) + geom_bar(aes(y = housing))
```



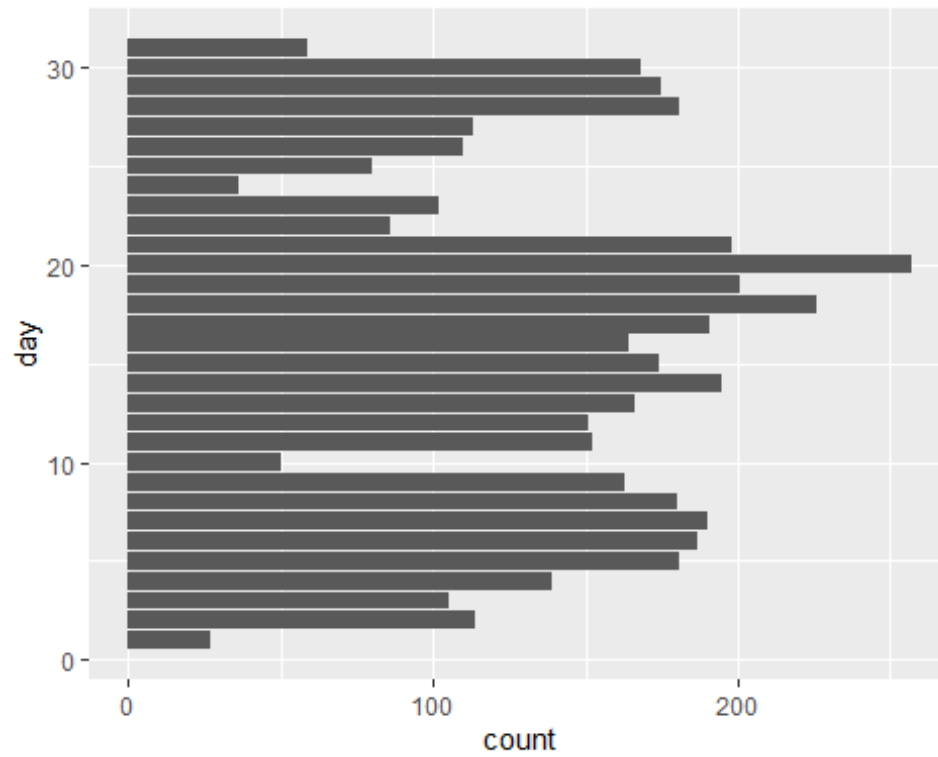
```
#plot loan row  
ggplot(data) + geom_bar(aes(y = loan))
```



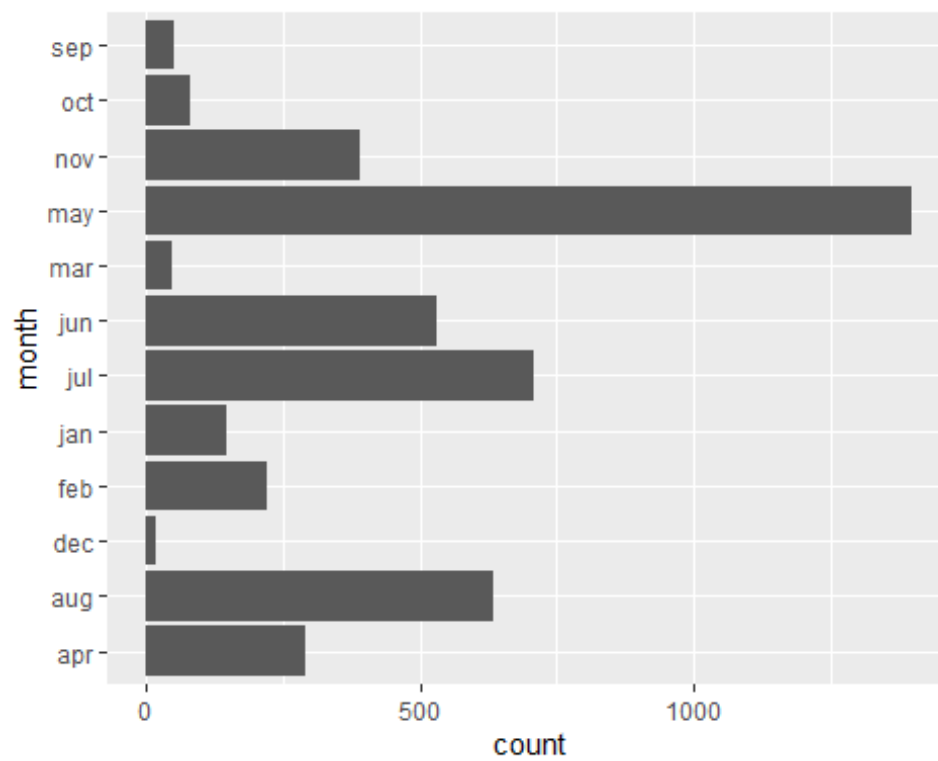
```
#plot contact row  
ggplot(data) + geom_bar(aes(y = contact))
```



```
#plot day row  
ggplot(data) + geom_bar(aes(y = day))
```

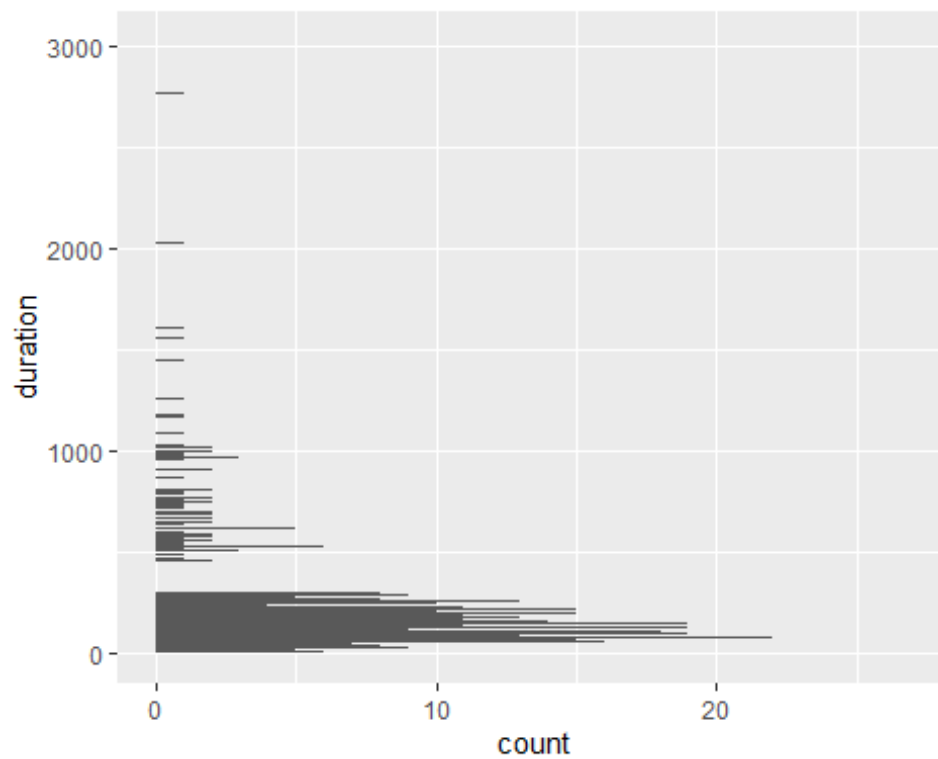


```
#plot month row  
ggplot(data) + geom_bar(aes(y = month))
```

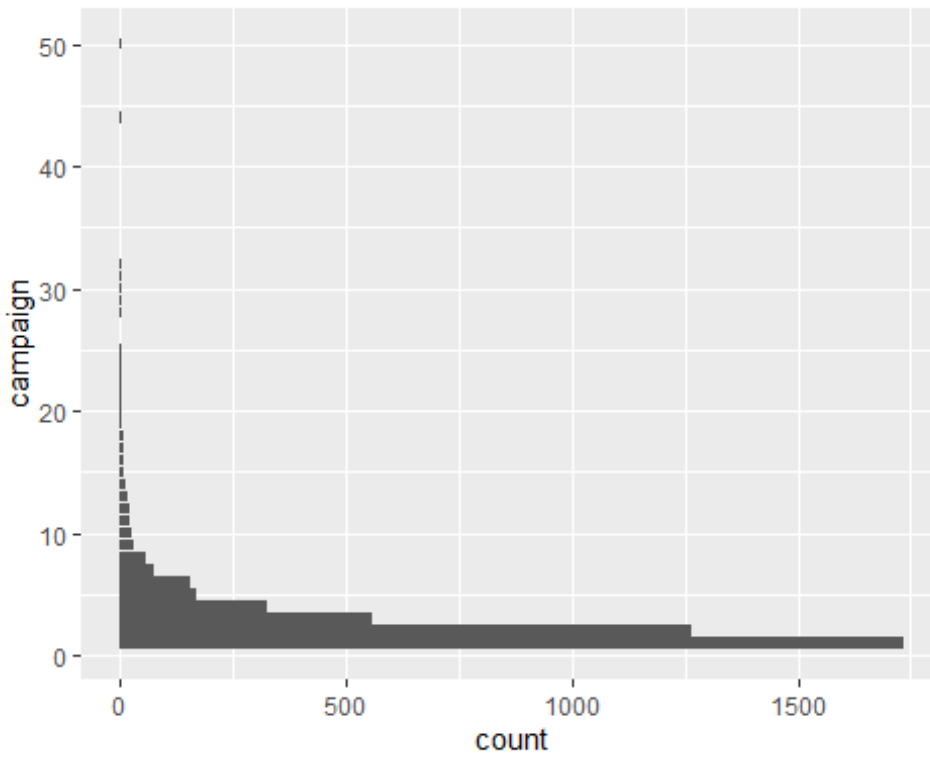




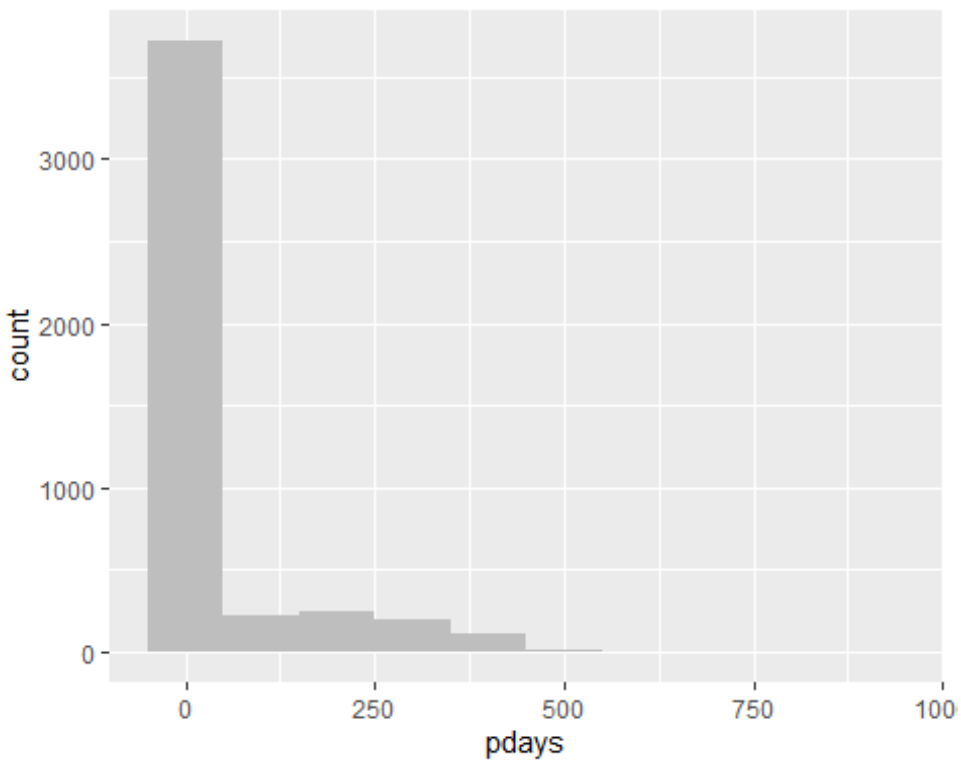
```
#plot duration row  
ggplot(data) + geom_bar(aes(y =duration))
```



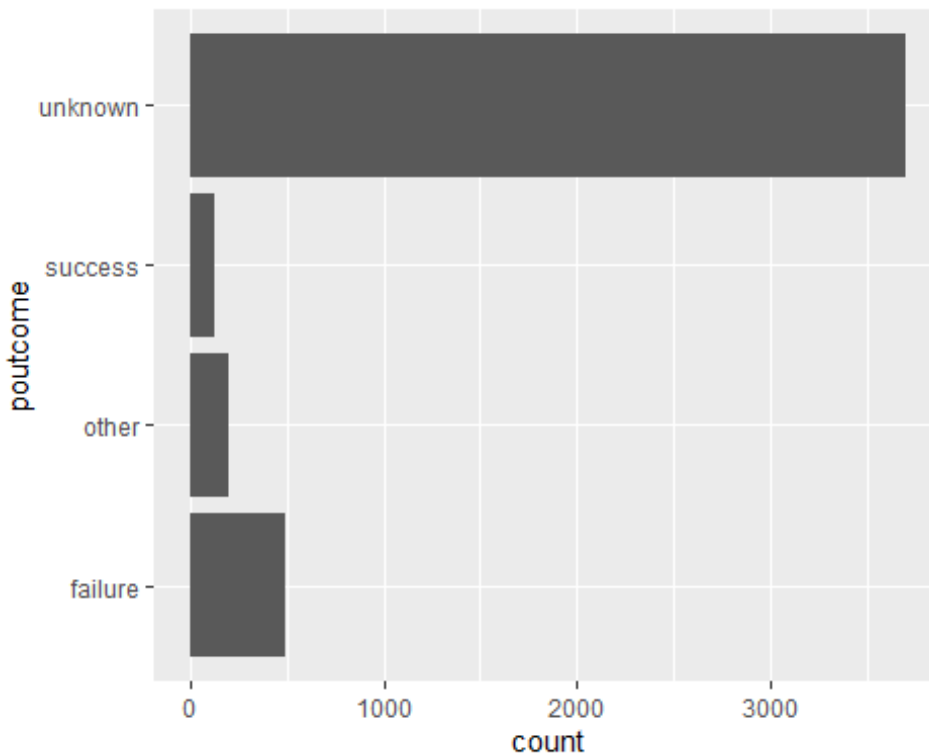
```
#plot campaign row  
ggplot(data) + geom_bar(aes(y =campaign))
```



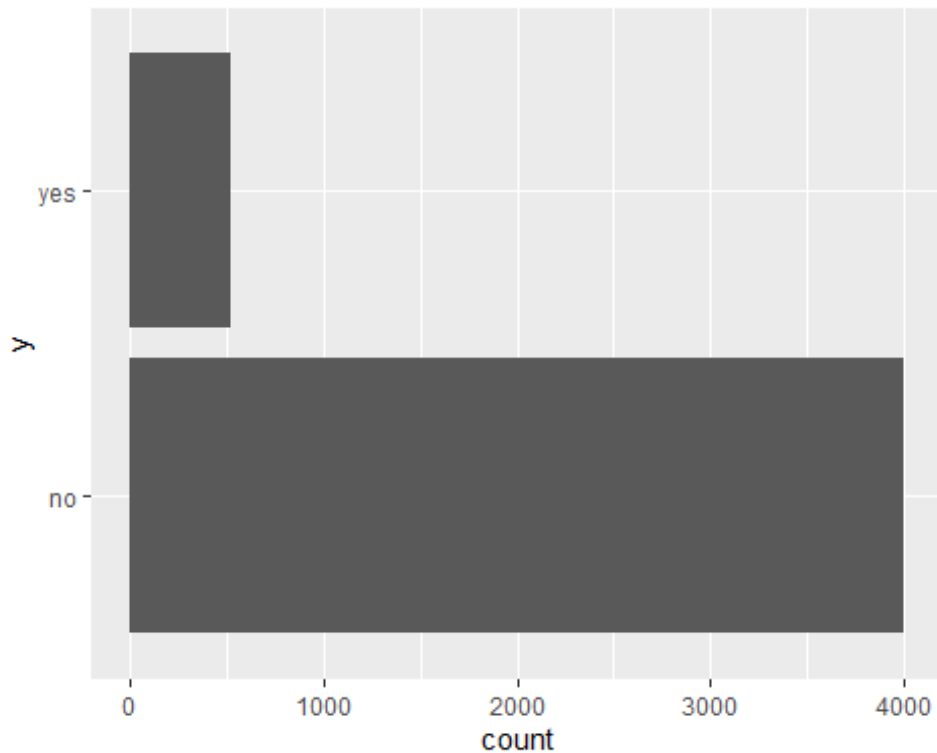
```
#for pdays row
ggplot(data, aes(x = pdays)) + geom_histogram(fill="grey",
position="dodge",binwidth=100)
```



```
data$pdays[data$pdays==-1] = mean(data$pdays)
#plot poutcome row
ggplot(data) + geom_bar(aes(y = poutcome))
```



```
#As poutcome also mostly has unknown only, so we remove it to prevent
overfitting from occuring
data = data[,-c(15)]
#plot y row
ggplot(data) + geom_bar(aes(y =y))
```



```
#for balance row
data$balance[data$balance==0] = mean(data$balance)
```

```
data <- transform(
  data,
  age = age,
  job = as.integer(factor(job)),
  marital = as.integer(factor(marital)),
  education=as.integer(factor(education)),
  balance = balance,
  housing = as.integer(factor(housing)),
  loan = as.integer(factor(loan)),
  contact = as.integer(factor(contact)),
  day = day,
  month = as.integer(factor(month)),
  duration = duration,
  campaign = campaign,
  pdays = pdays
)
sapply(data, class)
```

```
##      age      job      marital  education  balance  housing
## "integer" "integer" "integer"  "integer" "numeric" "integer"
##      loan      contact      day      month      duration  campaign
## "integer" "integer"  "integer"  "integer"  "integer"  "integer"
##      pdays      previous      y
## "numeric" "integer" "character"
```

```

#train test split
split = sample.split(data$y, SplitRatio=0.7)
training = subset(data, split==TRUE)
test = subset(data, split==FALSE)

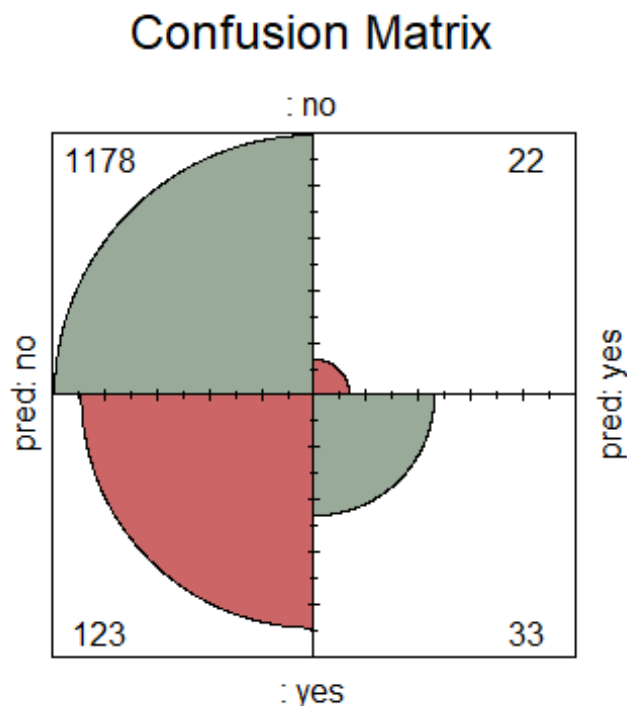
#scaling the data
training[,c(1:14)] = scale(training[,c(1:14)])

#random forest
rf <- randomForest(
  as.factor(y) ~ .,
  data=training,
)

test[,c(1:14)] = scale(test[,c(1:14)])

pred = predict(rf, newdata=test[,c(1:14)])
cm = table(test[,15], pred)
#confusion matrix
fourfoldplot(cm, color = c("#CC6666", "#99AA99"), conf.level = 0, margin = 1,
main = "Confusion Matrix")

```



```
res = confusionMatrix(cm)
```

```
print(res)
```

```

## Confusion Matrix and Statistics
##
##      pred
##      no  yes
## no 1178  22
## yes 123   33
##
##              Accuracy : 0.8931
##              95% CI : (0.8754, 0.909)
##      No Information Rate : 0.9594
##      P-Value [Acc > NIR] : 1
##
##              Kappa : 0.269
##
##  Mcnemar's Test P-Value : <2e-16
##
##              Sensitivity : 0.9055
##              Specificity : 0.6000
##              Pos Pred Value : 0.9817
##              Neg Pred Value : 0.2115
##              Prevalence : 0.9594
##              Detection Rate : 0.8687
##      Detection Prevalence : 0.8850
##      Balanced Accuracy : 0.7527
##
##      'Positive' Class : no
##

Precisionvalue = res$byClass["Pos Pred Value"]
print(Precisionvalue)

## Pos Pred Value
##      0.9816667

Accuracy=res$overall["Accuracy"]
print(Accuracy)

## Accuracy
## 0.8930678

```