CrimeAnalysis

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```
# 9.3 Crime data Question
# a. Fit a regression model
library(MASS)
# in page # 159, its clearly given the algorithm to use the MontoCarlo simulation
# using the code discussed by the professor in the class
UsCrimeData<-UScrime
# assining the values from the data to the variables or selection the relevant columns from the
  US crime data
UsCrimeResultantY<-UsCrimeData$y
CrimeDataToAnalyze<-UsCrimeData[,-16]</pre>
CrimeDataToAnalyze<-cbind(rep(1,47),CrimeDataToAnalyze)</pre>
CrimeDataToAnalyze<-as.matrix(CrimeDataToAnalyze)</pre>
# loaded all the required data above
# now lets initialize and set all the parameters required to start the iterative MC process
# initialize as shown in the text book
nu0<-2
s20=1
SigmaSquare=1
SamplingIters<-10000
Beta<-S2<-NULL
xtxinv<-solve(t(CrimeDataToAnalyze)%*%CrimeDataToAnalyze)
linearModelingResult<-lm(UsCrimeResultantY~CrimeDataToAnalyze[,-1])</pre>
# using code shared in text book and by Professor
SSRg<-t(UsCrimeResultantY)%*%UsCrimeResultantY-length(UsCrimeData$y)/(length(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeData$y)+1)*t(UsCrimeDa
CrimeResultantY)%*%predict(linearModelingResult)
for(i in 1:SamplingIters){
    s2<-1/rgamma(n=1,nu0+length(UsCrimeData$y)/2,(nu0*s20-SigmaSquare+SSRg)/2)</pre>
    ValueOfbeta<-mvrnorm(n=1,mu=length(UsCrimeData$y)/(length(UsCrimeData$y)+1)*linearModelingResu
lt$coefficients,length(UsCrimeData$y)/(length(UsCrimeData$y)+1)*s2*xtxinv)
    Beta<-rbind(Beta, ValueOfbeta)</pre>
    S2<-rbind(S2,s2)
}
###########################
## Now working on the posterior samples
posteriorsample<-data.frame(Beta,S2)</pre>
colnames(posteriorsample)<-c("Beta0", "Beta1", "Beta2", "Beta3", "Beta4", "Beta5", "Beta6", "Beta6", "Beta7", "Be
ta8", "Beta9", "Beta10", "Beta11", "Beta12", "Beta13", "Beta14", "Beta15", "Sigma2")
posteriorMean<-colMeans(posteriorsample)</pre>
quantiles<-matrix(NA,15,2)
quantiles<-apply(posteriorsample,2,function(CrimeDataToAnalyze)quantile(CrimeDataToAnalyze,c(0.0
25,0.975)))
print(quantiles)
```

```
##
             Beta0
                       Beta1
                                  Beta2
                                           Beta3
                                                     Beta4
                                                               Beta5
## 2.5% -9084.060 0.2054869 -289.0855 5.722993 -2.093905 -34.50221
## 97.5% -2532.647 16.9374886
                             293.5188 30.745987 40.237493 12.59582
             Beta6
                       Beta7
                                Beta8
                                           Beta9
                                                     Beta10
                                                                Beta11
## 2.5%
       -3.578822 -2.433715 -3.330872 -0.9194859 -14.067121 -0.2237992
## 97.5% 2.284444 5.684448 1.862107 1.7004010
                                                   2.705511 32.7217595
##
            Beta12
                     Beta13
                                Beta14
                                          Beta15
                                                   Sigma2
## 2.5%
        -1.186508 2.323024 -9386.1591 -17.65502 31238.28
## 97.5% 2.988666 11.414119 -299.5392 10.42213 68425.54
```

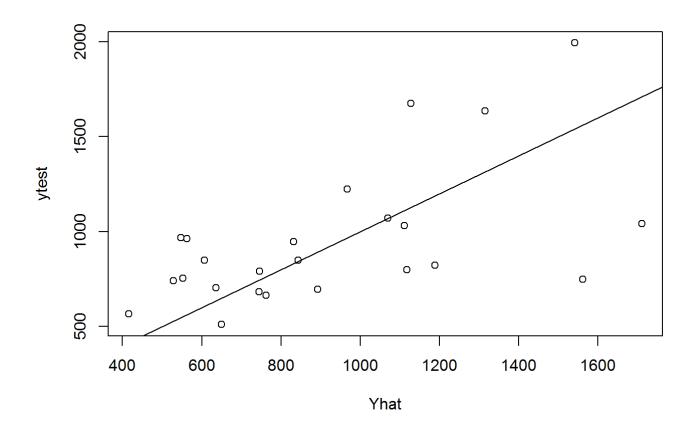
print(posteriorMean)

```
##
           Beta0
                          Beta1
                                         Beta2
                                                       Beta3
                                                                      Beta4
   -5842.3570099
                      8.5544937
                                   -0.4310164
                                                  18.4263013
                                                                 19.0548710
##
##
                          Beta6
                                                       Beta8
                                                                      Beta9
           Beta5
                                         Beta7
##
     -10.9191383
                     -0.6434110
                                    1.7072705
                                                  -0.6974008
                                                                  0.4066948
##
          Beta10
                         Beta11
                                                      Beta13
                                                                     Beta14
                                       Beta12
      -5.7003581
                     16.3760101
                                    0.9307654
                                                   6.8982190 -4795.8793622
##
##
          Beta15
                         Sigma2
##
      -3.5105993 46328.6092556
```

coffecients<-linearModelingResult\$coefficients
print(summary(linearModelingResult))</pre>

```
##
## Call:
## lm(formula = UsCrimeResultantY ~ CrimeDataToAnalyze[, -1])
## Residuals:
                1Q Median
##
       Min
                                3Q
                                       Max
## -395.74 -98.09
                    -6.69 112.99 512.67
##
## Coefficients:
##
                                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                -5984.2876 1628.3184 -3.675 0.000893 ***
## CrimeDataToAnalyze[, -1]M
                                    8.7830
                                               4.1714
                                                       2.106 0.043443 *
## CrimeDataToAnalyze[, -1]So
                                   -3.8035
                                             148.7551 -0.026 0.979765
                                                      3.033 0.004861 **
## CrimeDataToAnalyze[, -1]Ed
                                   18.8324
                                               6.2088
                                   19.2804
## CrimeDataToAnalyze[, -1]Po1
                                              10.6110
                                                       1.817 0.078892 .
## CrimeDataToAnalyze[, -1]Po2
                                              11.7478 -0.931 0.358830
                                  -10.9422
## CrimeDataToAnalyze[, -1]LF
                                               1.4697 -0.452 0.654654
                                   -0.6638
## CrimeDataToAnalyze[, -1]M.F
                                    1.7407
                                               2.0354
                                                      0.855 0.398995
## CrimeDataToAnalyze[, -1]Pop
                                               1.2896
                                                      -0.568 0.573845
                                   -0.7330
## CrimeDataToAnalyze[, -1]NW
                                    0.4204
                                               0.6481
                                                        0.649 0.521279
## CrimeDataToAnalyze[, -1]U1
                                   -5.8271
                                               4.2103 -1.384 0.176238
## CrimeDataToAnalyze[, -1]U2
                                   16.7800
                                               8.2336
                                                       2.038 0.050161 .
## CrimeDataToAnalyze[, -1]GDP
                                    0.9617
                                                      0.928 0.360754
                                               1.0367
## CrimeDataToAnalyze[, -1]Ineq
                                    7.0672
                                               2.2717
                                                        3.111 0.003983 **
## CrimeDataToAnalyze[, -1]Prob -4855.2658
                                            2272.3746 -2.137 0.040627 *
## CrimeDataToAnalyze[, -1]Time
                                                      -0.486 0.630708
                                   -3.4790
                                               7.1653
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 209.1 on 31 degrees of freedom
## Multiple R-squared: 0.8031, Adjusted R-squared: 0.7078
## F-statistic: 8.429 on 15 and 31 DF, p-value: 3.539e-07
```

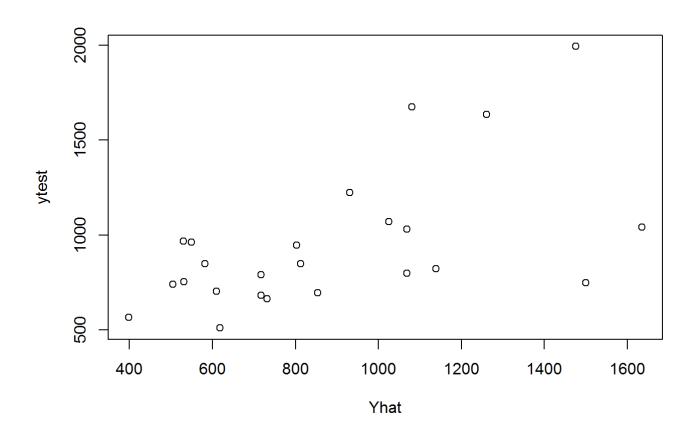
```
####################
#Question part 2 - b - 1
SizeOfSamlple<-0.5*length(UsCrimeData$y)</pre>
## set the seed to make your partition reproductible
set.seed(145)
trainDataindexValues <- sample(seq_len(length(UsCrimeData$y)), size = SizeOfSamlple)</pre>
train <- UsCrimeData[trainDataindexValues, ]</pre>
test <- UsCrimeData[-trainDataindexValues, ]</pre>
trainValueOfY<-as.matrix(train$y)</pre>
Xtrain<-as.matrix(train[,-16])</pre>
ytest<-as.matrix(test$y)</pre>
Xtest<-as.matrix(test[,-16])</pre>
Xtest<-cbind(rep(1,nrow(Xtest)),Xtest)</pre>
linear_model<-lm(trainValueOfY~Xtrain)</pre>
B<-linear_model$coefficients
Yhat<-Xtest%*%B
plot(Yhat,ytest)
abline(0,1)
```



```
perror<-(1/length(ytest))*sum((ytest-Yhat)^2)
print(perror)</pre>
```

[1] 108998

```
##Question part 2 - b - 2
train <- UsCrimeData[trainDataindexValues, ]</pre>
test <- UsCrimeData[-trainDataindexValues, ]</pre>
trainValueOfY<-as.matrix(train$y)</pre>
UsCrimeResultantY<-trainValueOfY
Xtrain<-as.matrix(train[,-16])</pre>
Xtrain<-cbind(rep(1,nrow(Xtrain)),Xtrain)</pre>
CrimeDataToAnalyze<-Xtrain
ytest<-as.matrix(test$y)</pre>
Xtest<-as.matrix(test[,-16])</pre>
Xtest<-cbind(rep(1,nrow(Xtest)),Xtest)</pre>
nu0<-2
s20 <- SigmaSquare<-1
SamplingIters<-10000
Beta<-S2<-NULL
n<-length(trainValueOfY)</pre>
xtxinv<-solve(t(CrimeDataToAnalyze)%*%CrimeDataToAnalyze)
linearModelingResult<-lm(UsCrimeResultantY~CrimeDataToAnalyze[,-1])</pre>
SSRg<-t(UsCrimeResultantY)%*%UsCrimeResultantY-length(trainValueOfY)/(length(trainValueOfY)+1)*t(U
CrimeResultantY)%*%predict(linearModelingResult)
for(i in 1:SamplingIters){
  s2<-1/rgamma(n=1,nu0+length(trainValueOfY)/2,(nu0*s20-SigmaSquare+SSRg)/2)</pre>
  ValueOfbeta<-mvrnorm(n=1,mu=length(trainValueOfY)/(length(trainValueOfY)+1)*linearModelingResu
lt$coefficients,length(trainValueOfY)/(length(trainValueOfY)+1)*s2*xtxinv)
  Beta<-rbind(Beta, ValueOfbeta)</pre>
  S2<-rbind(S2,s2)
}
posteriorsample<-data.frame(Beta)</pre>
colnames(posteriorsample)<-c("Beta0", "Beta1", "Beta2", "Beta3", "Beta4", "Beta5", "Beta6", "Beta6", "Beta7", "Be
ta8", "Beta9", "Beta10", "Beta11", "Beta12", "Beta13", "Beta14", "Beta15")
posteriorMean<-colMeans(posteriorsample)</pre>
posteriorMean<-as.matrix(posteriorMean)</pre>
Yhat<-Xtest%*%posteriorMean
plot(Yhat,ytest)
```



perror<-(1/length(ytest))*sum((ytest-Yhat)^2)
print(perror)</pre>

[1] 107600.3