

Project Linear Regression

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```
library(psych)
library(car)
```

```
##
## Attaching package: 'car'
```

```
## The following object is masked from 'package:psych':
##
##      logit
```

#1. Describe data of response variable and predictors in terms of key summary statistics like mean, mode, median, standard deviation, range, skewness and kurtosis. Show histogram and box plots also for each variables. [hint: describe command in R]
#Each variable to be explained in 30 words maximum.

```
grades<-read.csv(file.choose())
dim(grades)
```

```
## [1] 105 22
```

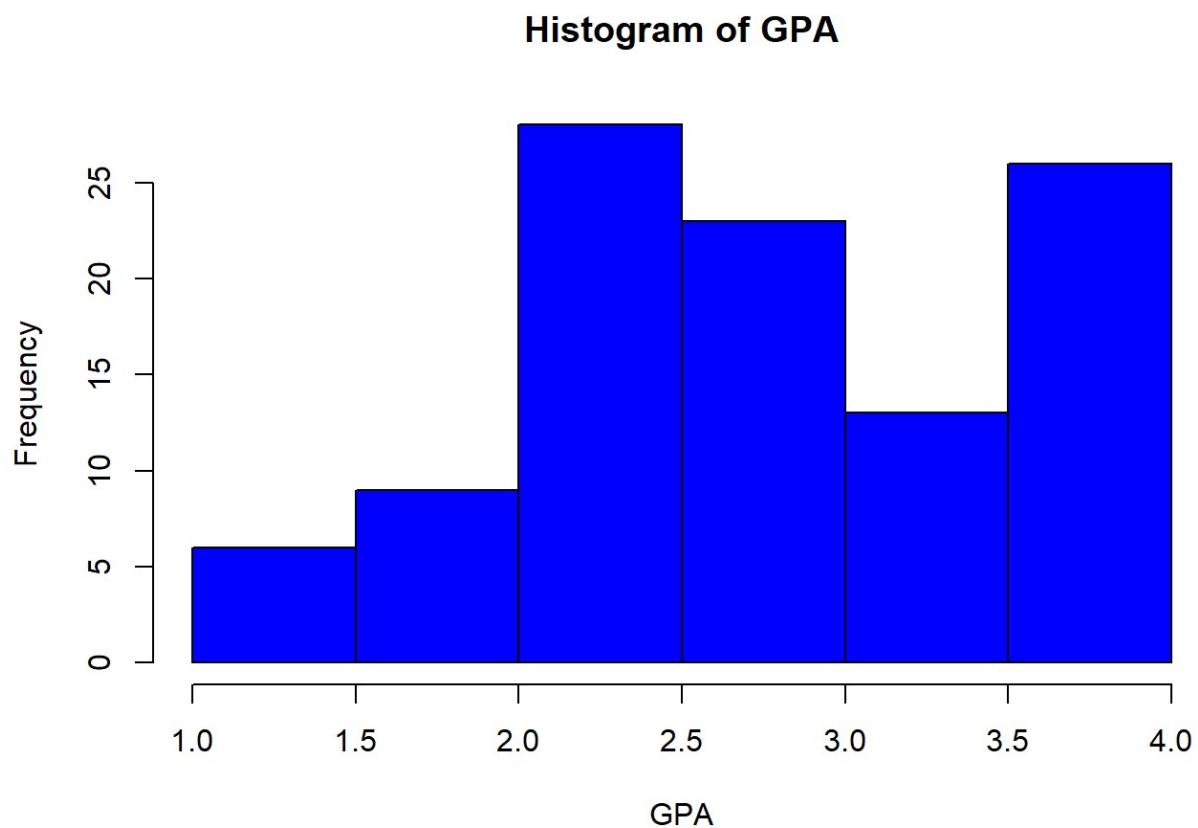
```
colnames(grades)
```

```
## [1] "Sr_No"      "id"        "lastname"   "firstname"  "gender"
## [6] "ethnicity"   "year"       "lowup"      "section"    "gpa"
## [11] "extrc"       "review"     "quiz1"      "quiz2"      "quiz3"
## [16] "quiz4"       "quiz5"      "final"      "total"      "percent"
## [21] "grade"       "passfail"
```

```
names(grades)
```

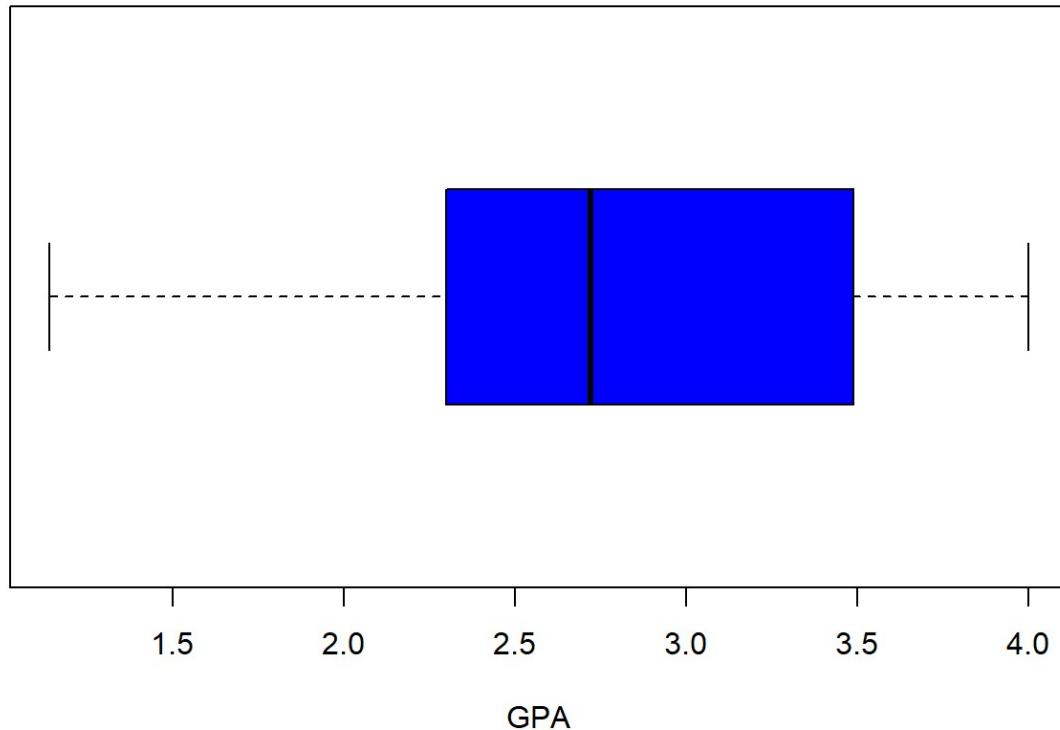
```
## [1] "Sr_No"      "id"        "lastname"   "firstname"  "gender"  
## [6] "ethnicity"   "year"       "lowup"      "section"    "gpa"  
## [11] "extrc"       "review"     "quiz1"      "quiz2"      "quiz3"  
## [16] "quiz4"       "quiz5"      "final"      "total"      "percent"  
## [21] "grade"       "passfail"
```

```
hist(grades$gpa, main = "Histogram of GPA", xlab = "GPA", ylab = "Frequency", col = "Blue")
```



```
boxplot(grades$gpa, main = "Box plot of GPA", xlab = "GPA", col = "Blue", horizontal = T)
```

Box plot of GPA



```
stem(grades$gpa)
```

```
##  
##      The decimal point is 1 digit(s) to the left of the |  
##  
## 10 | 48  
## 12 | 443  
## 14 | 0  
## 16 | 1677  
## 18 | 04515  
## 20 | 23919  
## 22 | 1225780123445788  
## 24 | 03356671446667  
## 26 | 1623477  
## 28 | 004480266  
## 30 | 1256379  
## 32 | 85  
## 34 | 2599334778  
## 36 | 460  
## 38 | 440000000000058  
## 40 | 0
```

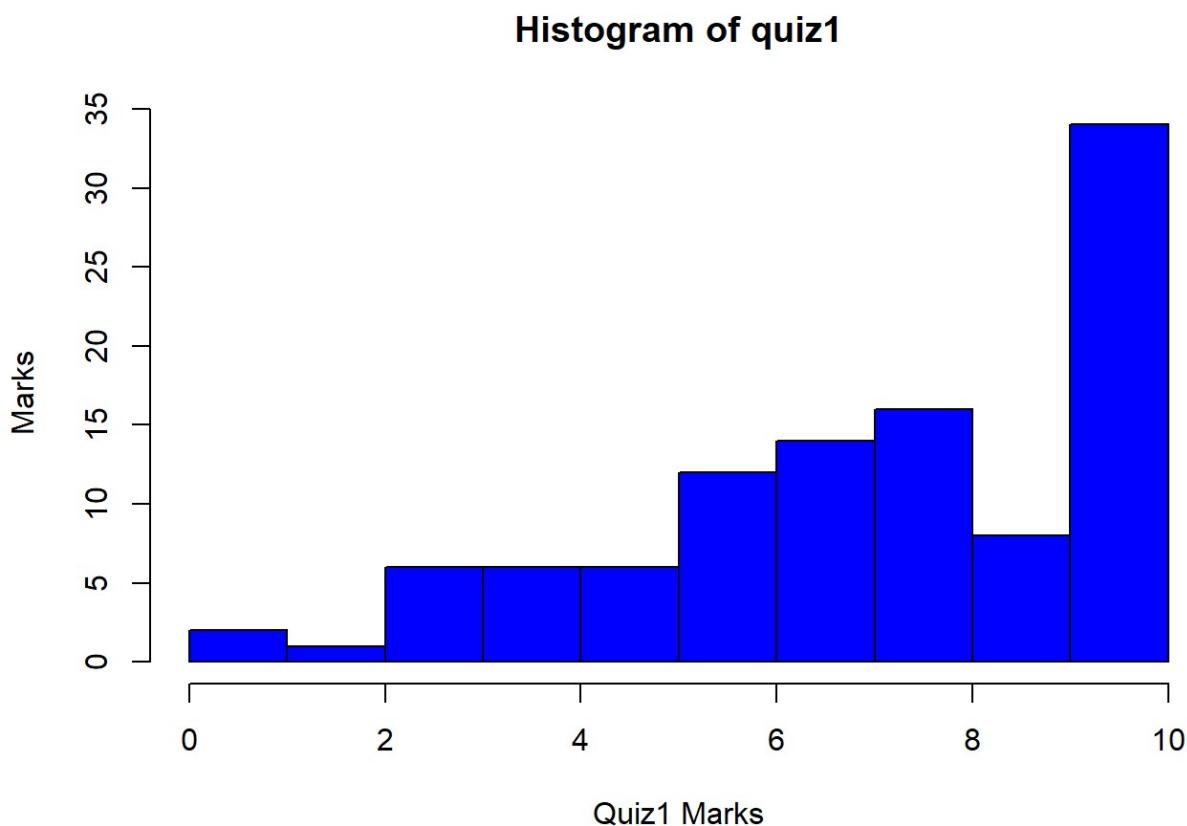
```
summary(grades$quiz1)
```

```
##      Min. 1st Qu. Median     Mean 3rd Qu.    Max.
## 0.000   6.000  8.000  7.467 10.000 10.000
```

```
describe(grades$quiz1)
```

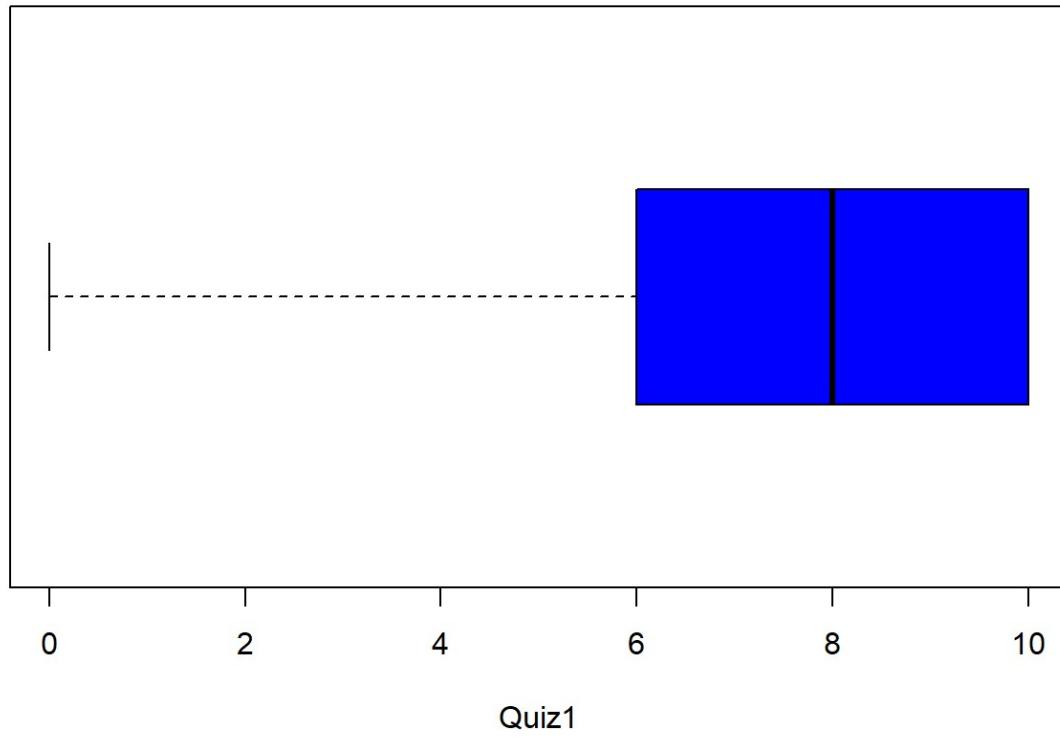
```
##    vars   n  mean   sd median trimmed  mad min max range skew kurtosis
## X1     1 105 7.47 2.48      8    7.76 2.97   0  10     10 -0.83     0.04
##          se
## X1  0.24
```

```
hist(grades$quiz1, main = "Histogram of quiz1", xlab = "Quiz1 Marks", ylab = "Marks",
col = "Blue")
```



```
boxplot(grades$quiz1, main = "Box plot of Quiz1", xlab = "Quiz1", col = "Blue", horizontal = T)
```

Box plot of Quiz1



```
stem(grades$quiz1)
```

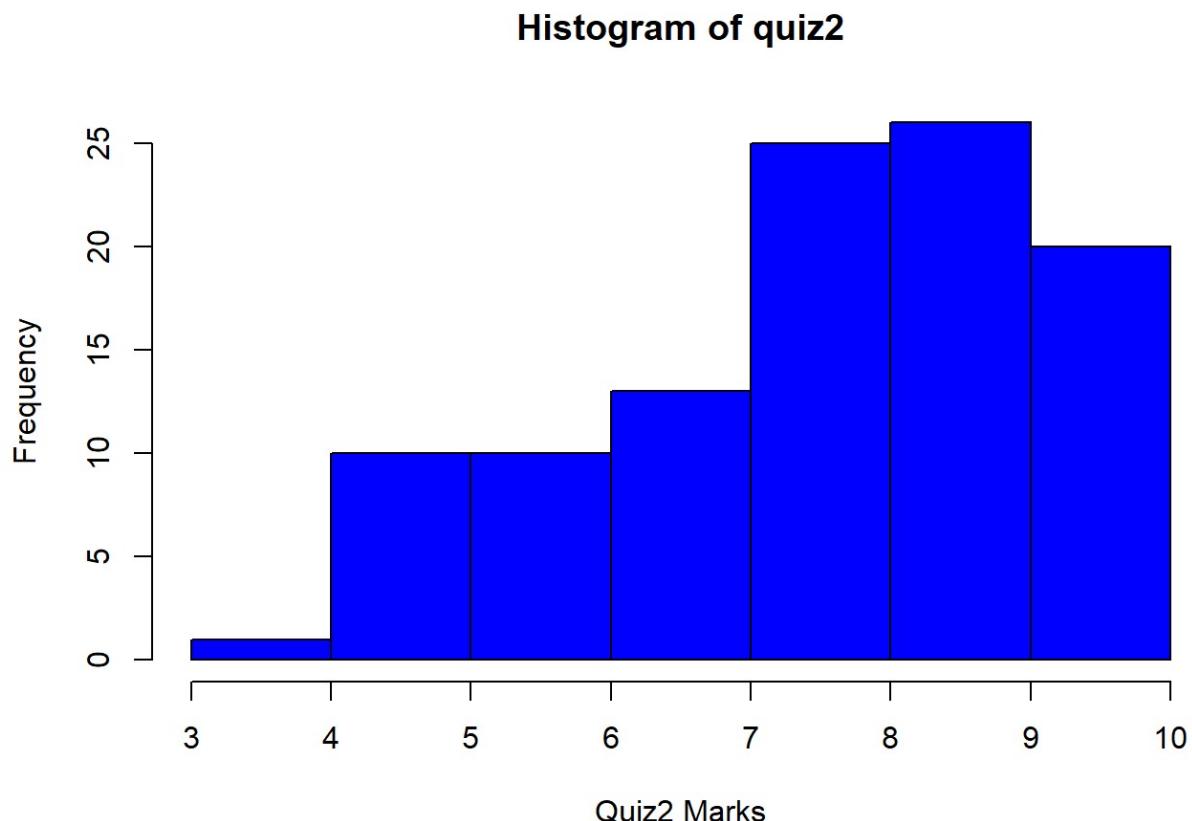
```
summary(grades$quiz2)
```

```
##      Min. 1st Qu. Median     Mean 3rd Qu.     Max.
##    3.000   7.000   8.000   7.981   9.000  10.000
```

```
describe(grades$quiz2)
```

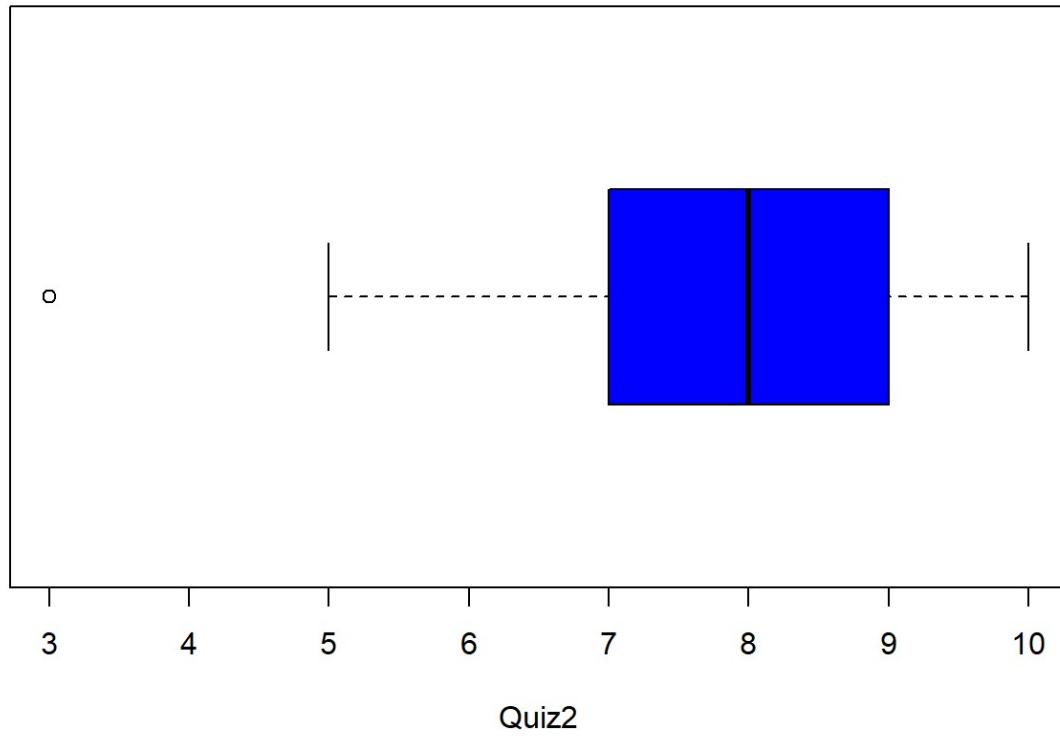
```
##      vars   n  mean   sd median trimmed  mad min max range skew kurtosis
## X1     1 105 7.98 1.62       8     8.12 1.48    3 10      7 -0.64    -0.35
##      se
## X1 0.16
```

```
hist(grades$quiz2, main = "Histogram of quiz2", xlab = "Quiz2 Marks", ylab = "Frequency", col = "Blue")
```



```
boxplot(grades$quiz2, main = "Box plot of Quiz2", xlab = "Quiz2", col = "Blue", horizontal = T)
```

Box plot of Quiz2



```
stem(grades$quiz2)
```

```
summary(grades$quiz3)
```

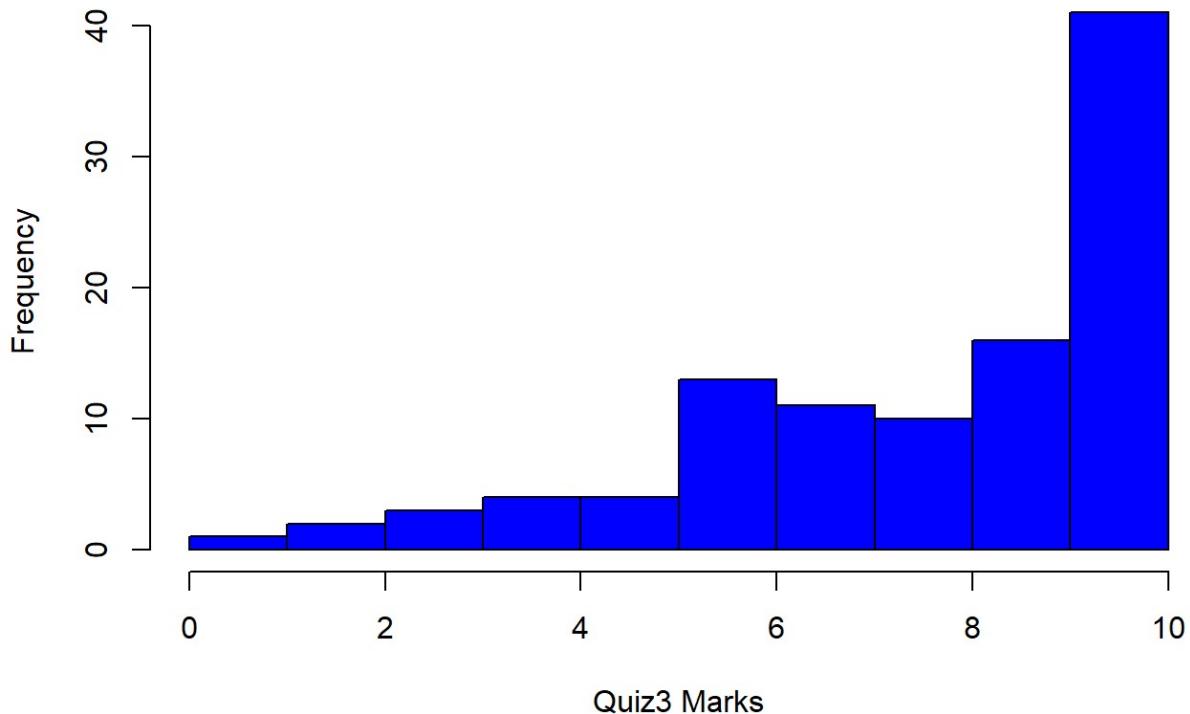
```
##      Min. 1st Qu. Median     Mean 3rd Qu.    Max.
## 0.000   6.000  9.000  7.981 10.000 10.000
```

```
describe(grades$quiz3)
```

```
##    vars   n  mean   sd median trimmed  mad min max range skew kurtosis   se
## X1     1 105 7.98 2.31      9     8.34 1.48    0   10     10 -1.1      0.59 0.23
```

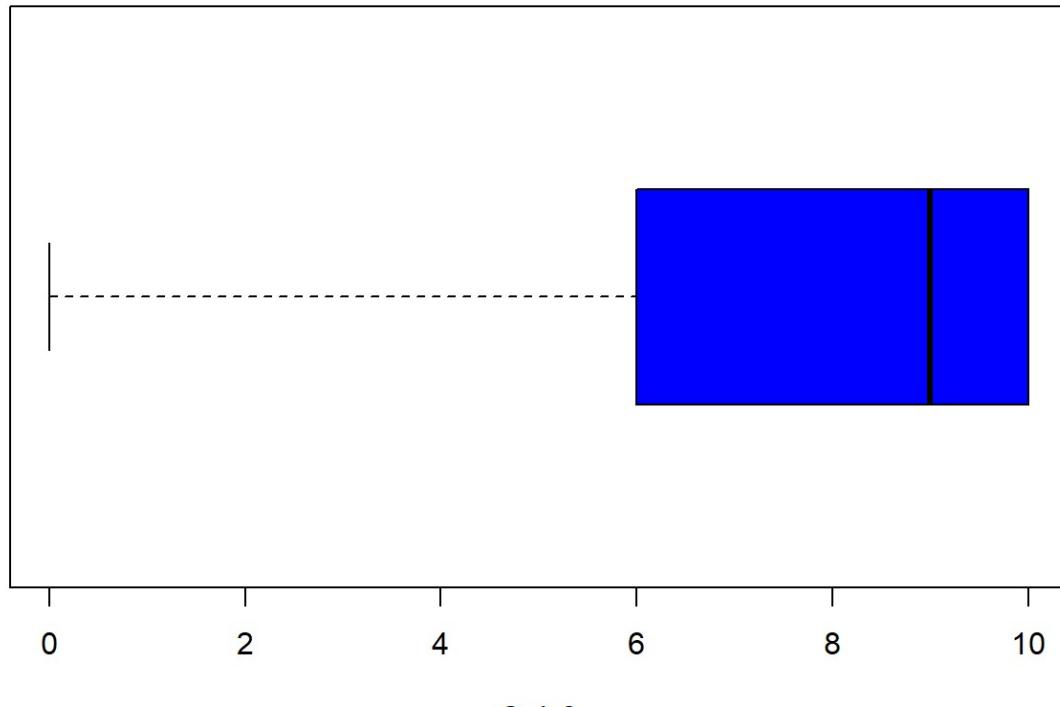
```
hist(grades$quiz3, main = "Histogram of quiz3", xlab = "Quiz3 Marks", ylab = "Frequency", col = "Blue")
```

Histogram of quiz3



```
boxplot(grades$quiz3, main = "Box plot of Quiz3", xlab = "Quiz3", col = "Blue", horizontal = T)
```

Box plot of Quiz3



```
stem(grades$quiz3)
```

```
summary(grades$quiz4)
```

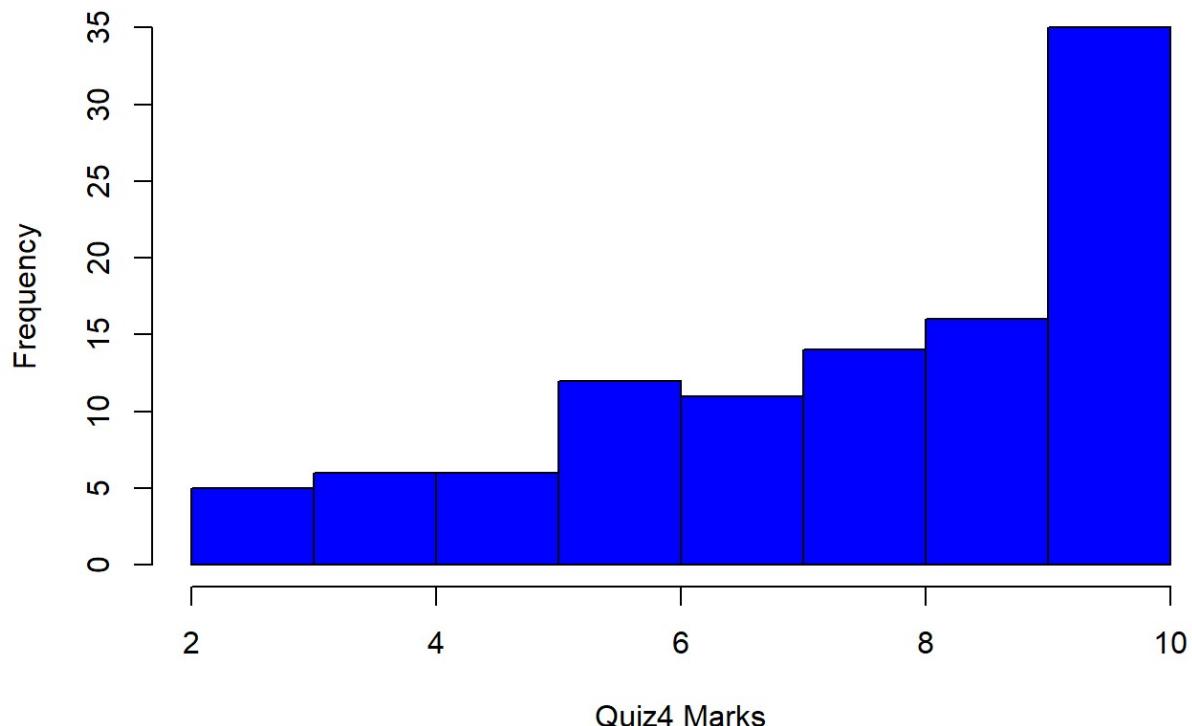
```
##      Min. 1st Qu. Median      Mean 3rd Qu.      Max.  
##      2.0     6.0     8.0     7.8    10.0    10.0
```

```
describe(grades$quiz4)
```

```
##      vars   n  mean   sd median trimmed  mad min max range skew kurtosis  
## X1     1 105  7.8 2.28       8     8.11 2.97    2 10       8 -0.89   -0.09  
##      se  
## X1 0.22
```

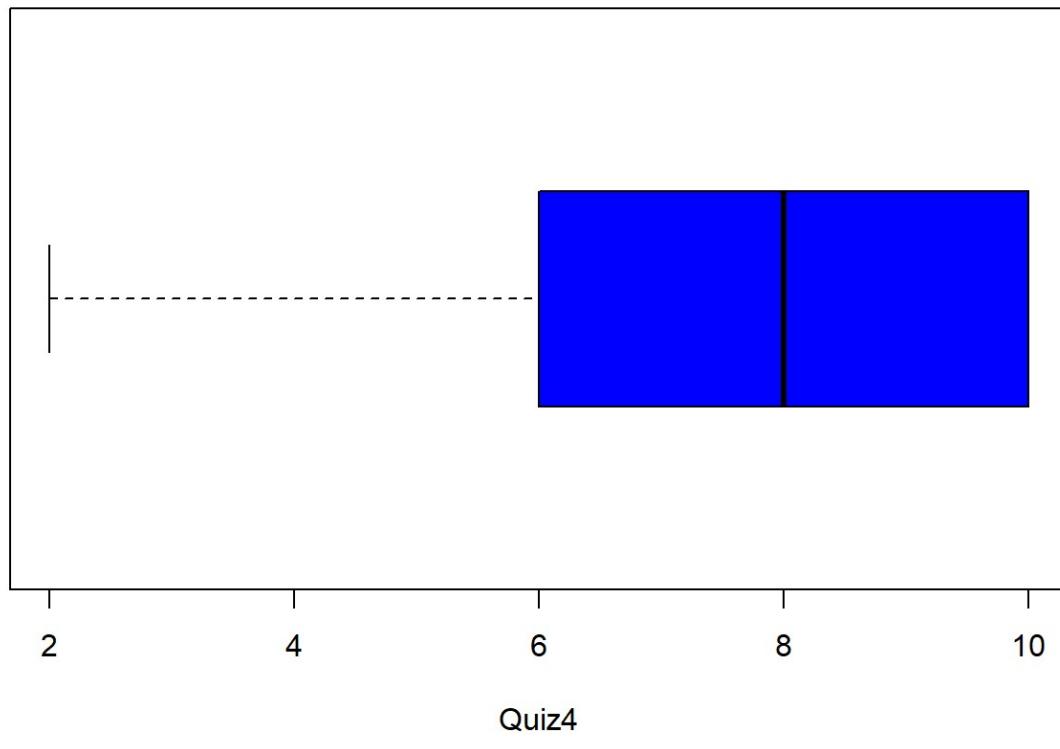
```
hist(grades$quiz4, main = "Histogram of quiz4", xlab = "Quiz4 Marks", ylab = "Frequency", col = "Blue")
```

Histogram of quiz4



```
boxplot(grades$quiz4, main = "Box plot of Quiz4", xlab = "Quiz4", col = "Blue", horizontal = T)
```

Box plot of Quiz4



```
stem(grades$quiz4)
```

```
summary(grades$quiz5)
```

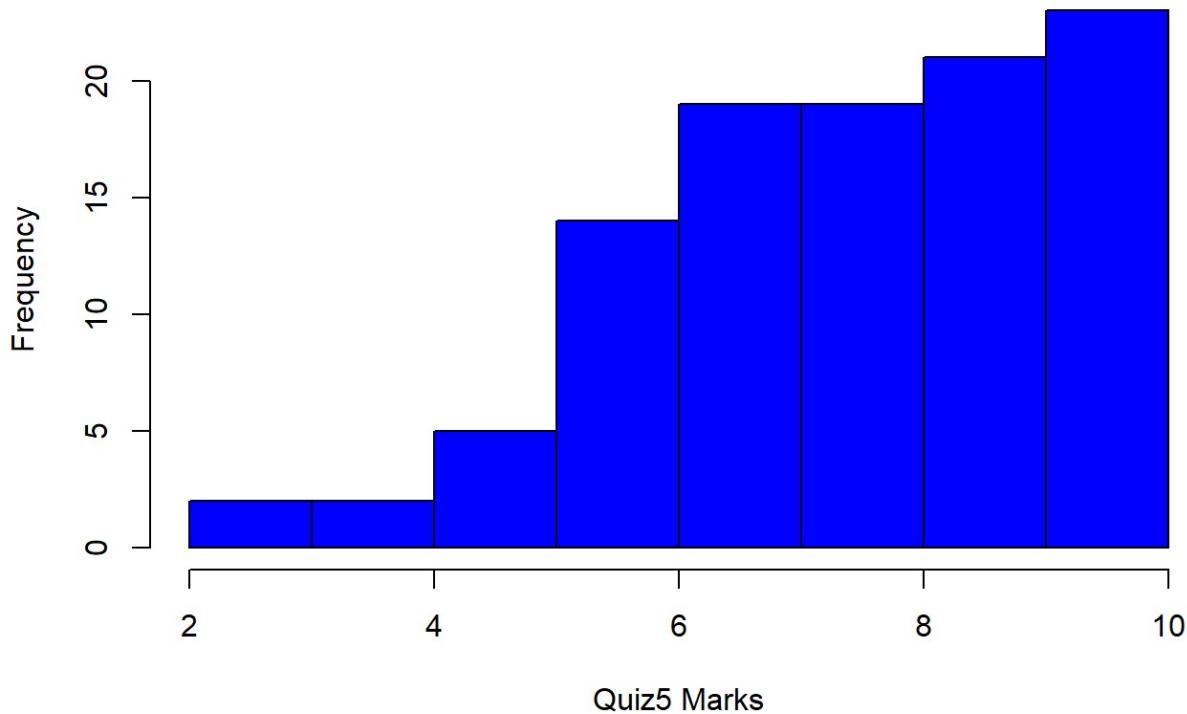
```
##      Min. 1st Qu. Median     Mean 3rd Qu.    Max.
##    2.000   7.000  8.000   7.867  9.000 10.000
```

```
describe(grades$quiz5)
```

```
##    vars   n  mean   sd median trimmed  mad min max range skew kurtosis
## X1     1 105 7.87 1.77      8     8.02 1.48   2  10      8 -0.69      0.16
##       se
## X1  0.17
```

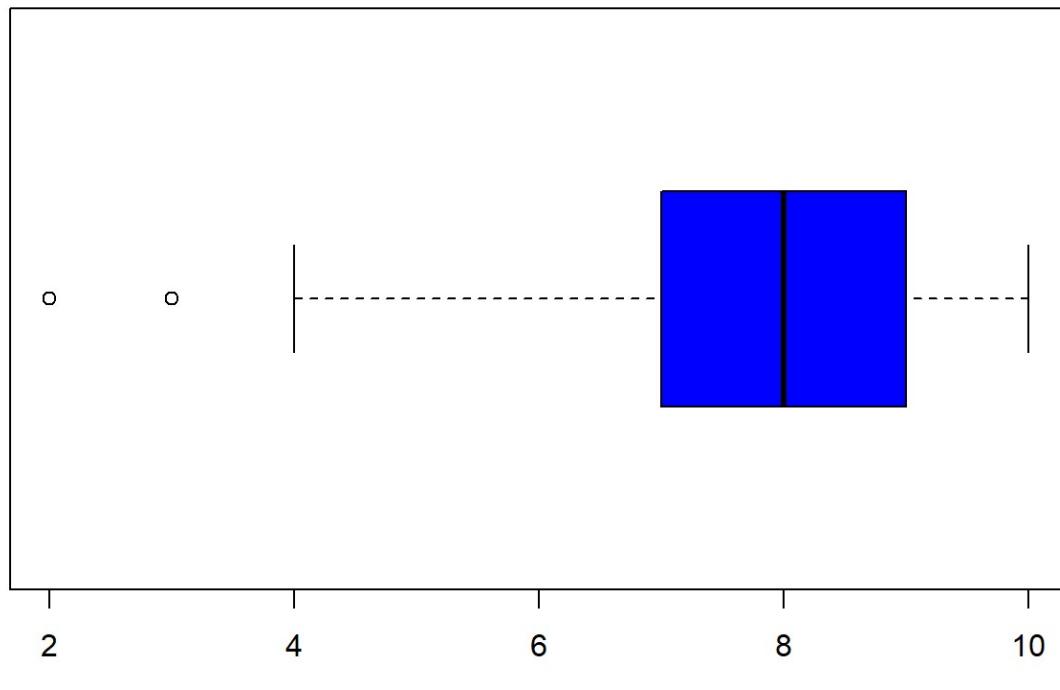
```
hist(grades$quiz5, main = "Histogram of quiz5", xlab = "Quiz5 Marks", ylab = "Frequency", col = "Blue")
```

Histogram of quiz5



```
boxplot(grades$quiz5, main = "Box plot of Quiz5", xlab = "Quiz5", col = "Blue", horizontal = T)
```

Box plot of Quiz5



```
stem(grades$quiz5)
```

```
##  
## The decimal point is at the  
##  
##      2 | 0  
##      2 |  
##      3 | 0  
##      3 |  
##      4 | 00  
##      4 |  
##      5 | 00000  
##      5 |  
##      6 | 00000000000000  
##      6 |  
##      7 | 00000000000000000000  
##      7 |  
##      8 | 00000000000000000000  
##      8 |  
##      9 | 00000000000000000000  
##      9 |  
##     10 | 00000000000000000000
```

```
summary(grades$final)
```

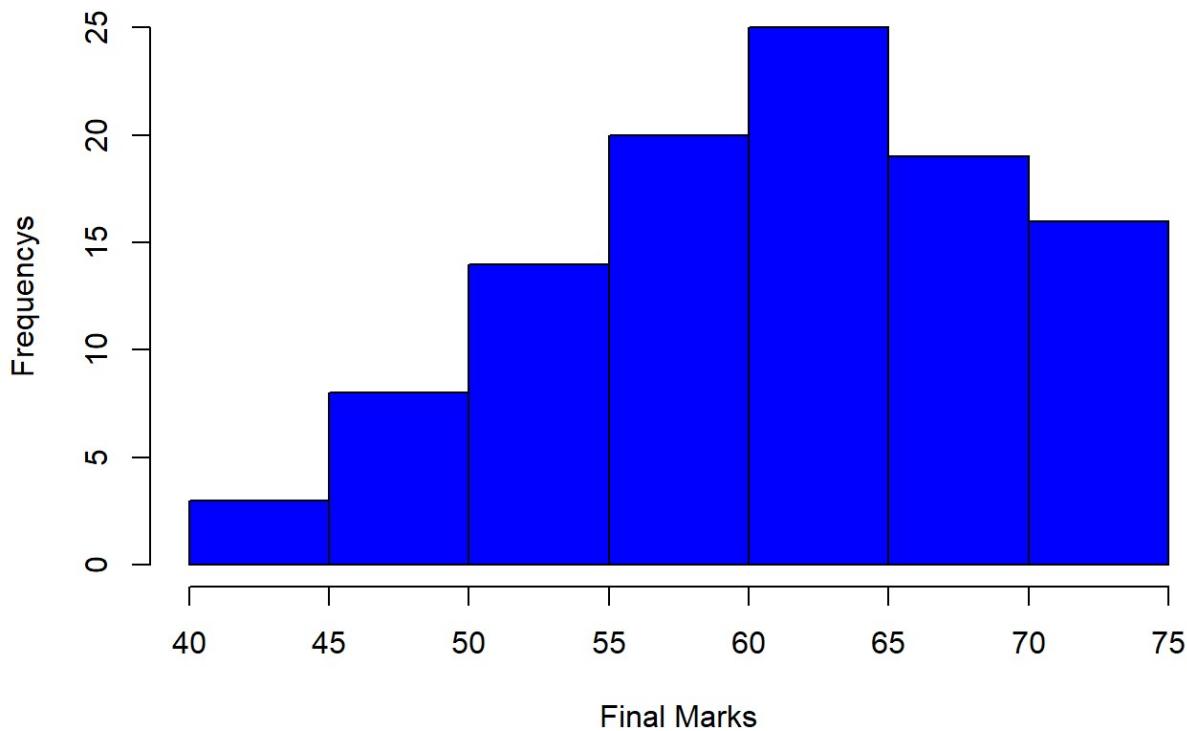
```
##      Min. 1st Qu. Median     Mean 3rd Qu.    Max.
## 40.00   57.00  62.00   61.48  68.00   75.00
```

```
describe(grades$final)
```

```
##    vars   n   mean    sd median trimmed mad min max range skew kurtosis
## X1     1 105 61.48 7.94      62   61.74 8.9   40   75     35 -0.33 -0.42
##          se
## X1 0.78
```

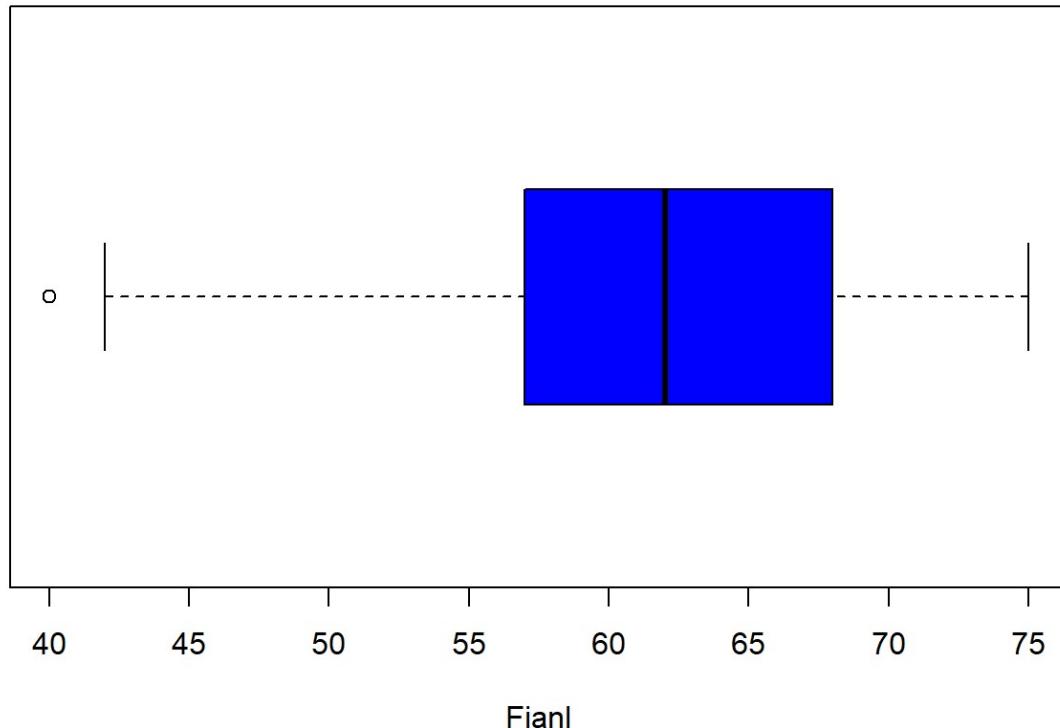
```
hist(grades$final, main = "Histogram of final", xlab = "Final Marks", ylab = "Frequencies", col = "Blue")
```

Histogram of final



```
boxplot(grades$final, main = "Box plot of Final", xlab = "Final", col = "Blue", horizontal = T)
```

Box plot of Final



```
stem(grades$final)
```

```
##  
## The decimal point is at the |  
##  
## 40 | 0  
## 42 | 00  
## 44 |  
## 46 |  
## 48 | 00000  
## 50 | 000  
## 52 | 000000000  
## 54 | 00000  
## 56 | 0000000  
## 58 | 000000  
## 60 | 000000000000  
## 62 | 00000000000000  
## 64 | 000000  
## 66 | 0000000  
## 68 | 0000000000  
## 70 | 000000  
## 72 | 00000  
## 74 | 0000000
```

```
summary(grades$total)
```

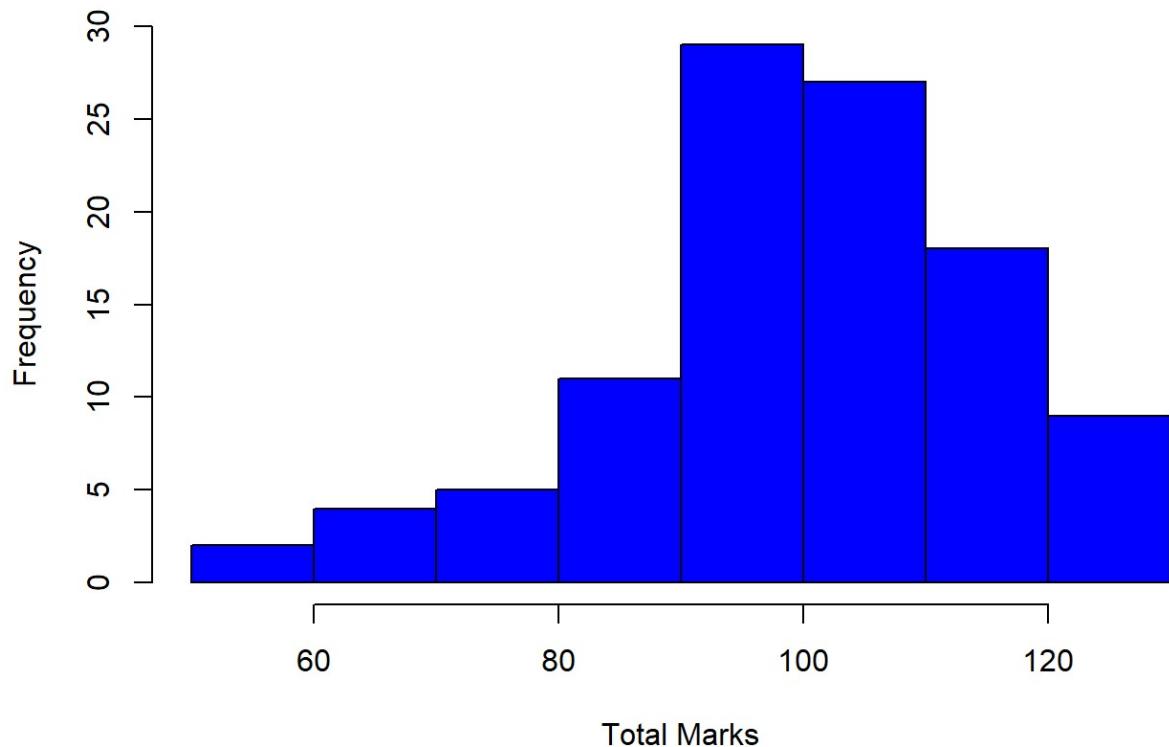
```
##   Min. 1st Qu. Median   Mean 3rd Qu.   Max.  
## 51.0    92.0   103.0  100.6   111.0  124.0
```

```
describe(grades$total)
```

```
##    vars   n   mean    sd median trimmed   mad min max range skew kurtosis  
## X1     1 105 100.57 15.3     103    101.8 13.34  51 124     73 -0.81      0.77  
##          se  
## X1 1.49
```

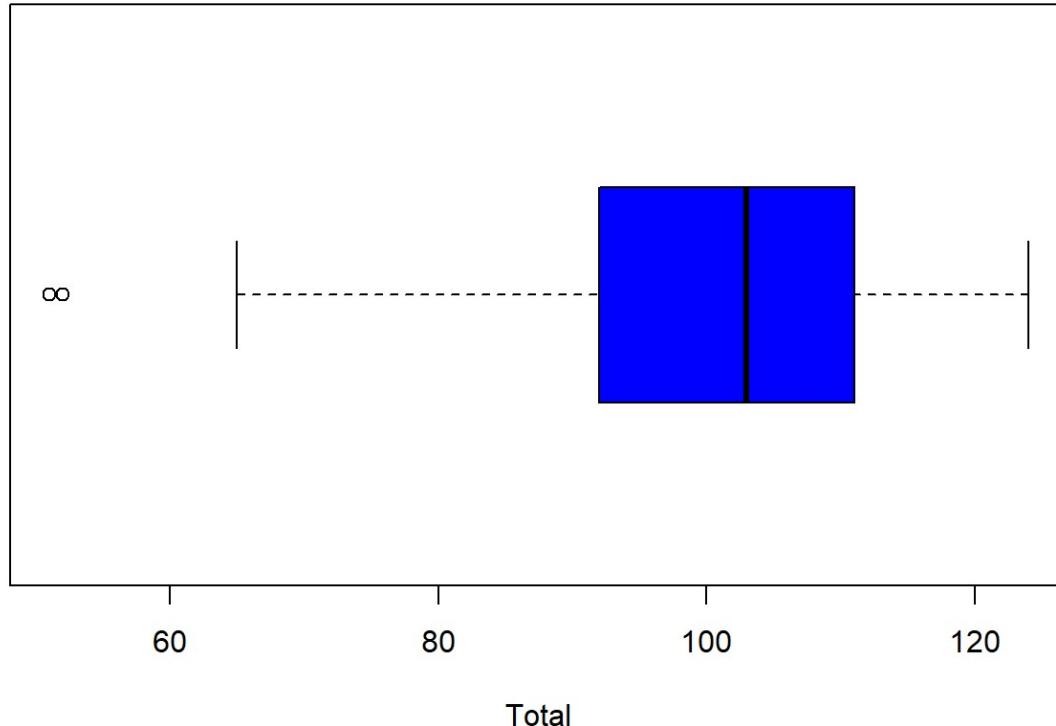
```
hist(grades$total, main = "Histogram of Total Marks", xlab = "Total Marks", ylab = "Frequency", col = "Blue")
```

Histogram of Total Marks



```
boxplot(grades$total, main = "Box plot of Total Marks", xlab = "Total", col = "Blue", horizontal = T)
```

Box plot of Total Marks



```
stem(grades$total)
```

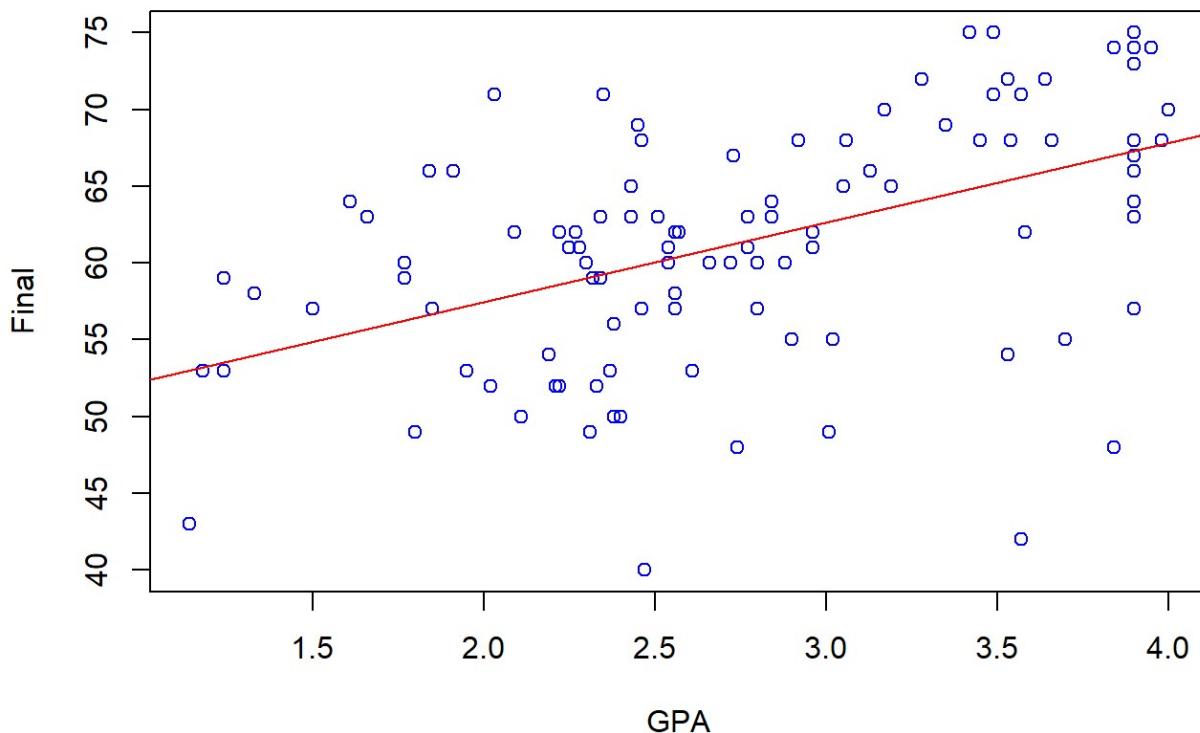
```
##  
##      The decimal point is 1 digit(s) to the right of the |  
##  
##      5 | 12  
##      5 |  
##      6 |  
##      6 | 56  
##      7 | 00  
##      7 | 5789  
##      8 | 01444  
##      8 | 7888  
##      9 | 00022222334  
##      9 | 5666677778888888999  
##      10 | 0133344  
##      10 | 556666777788888999999  
##      11 | 11123334  
##      11 | 5678888  
##      12 | 000122233344
```

```
#2. How predictor/s is related to response variable (final)? [hint: first plot scatter diagram followed by correlation test]
#Present diagram/s and correlations in the following space. Before diagrams explain relationship in 3 or 4 lines.
cor.test(grades$final, grades$gpa)
```

```
## 
## Pearson's product-moment correlation
##
## data: grades$final and grades$gpa
## t = 5.8291, df = 103, p-value = 6.44e-08
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.3387243 0.6296171
## sample estimates:
## cor
## 0.498055
```

```
plot(final~gpa, data = grades, main= "Scatter plot of GPA vs Final", col= 'blue', xlab = 'GPA', ylab="Final")
abline(lm(grades$final~grades$gpa), col="red")
```

Scatter plot of GPA vs Final

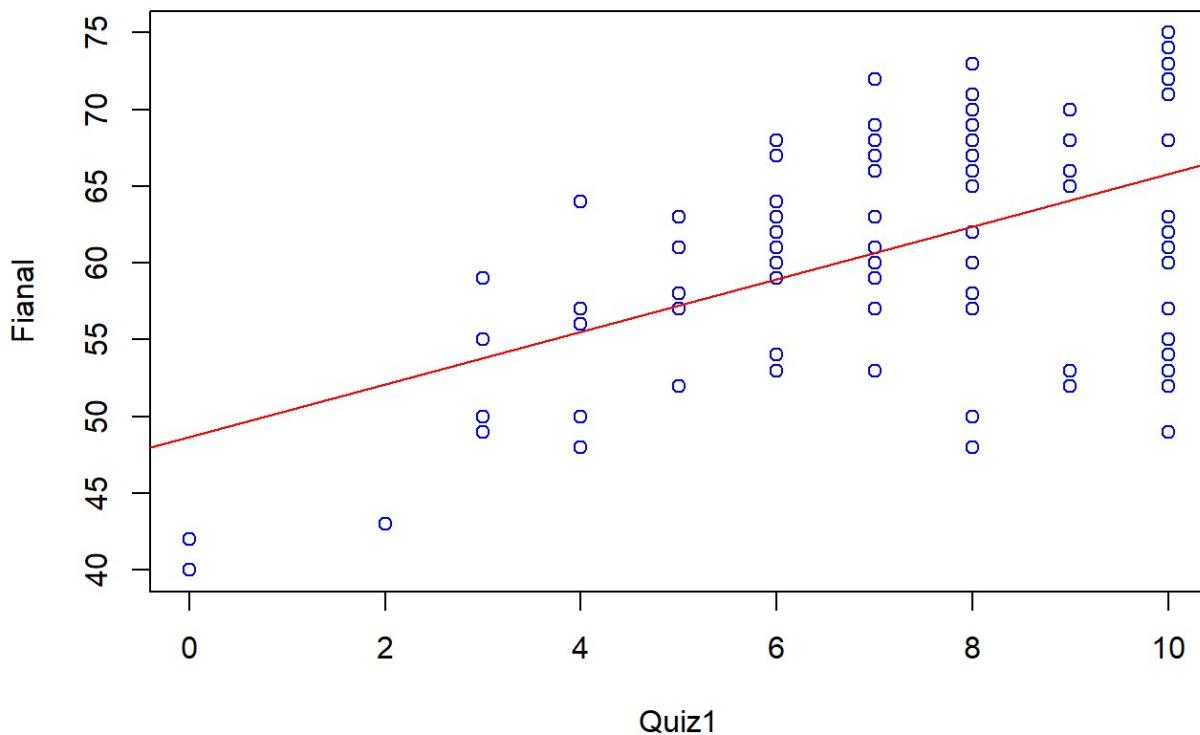


```
cor.test(grades$final, grades$quiz1)
```

```
##  
## Pearson's product-moment correlation  
##  
## data: grades$final and grades$quiz1  
## t = 6.428, df = 103, p-value = 4.094e-09  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
## 0.3826533 0.6591421  
## sample estimates:  
## cor  
## 0.5350754
```

```
plot(final~quiz1, data = grades, main= "Scatter plot of Quiz1 vs Final", col= 'blue', y  
lab= 'Final', xlab="Quiz1")  
abline(lm(grades$final~grades$quiz1), col="red")
```

Scatter plot of Quiz1 vs Final



```
cor.test(grades$final, grades$quiz2)
```

```

## 
## Pearson's product-moment correlation
## 
## data: grades$final and grades$quiz2
## t = 6.7162, df = 103, p-value = 1.047e-09
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.4028080 0.6724089
## sample estimates:
## cor
## 0.5518668

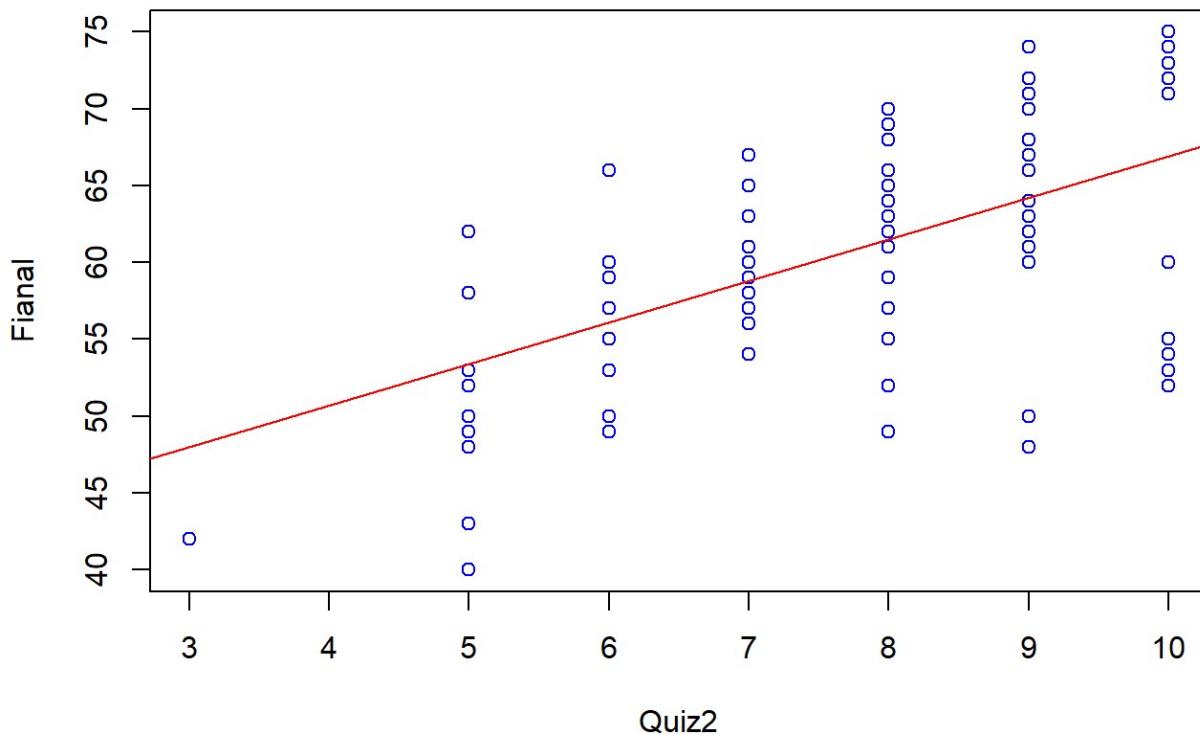
```

```

plot(final~quiz2, data = grades, main= "Scatter plot of Quiz2 vs Final", col= 'blue', y
lab= 'Final', xlab="Quiz2")
abline(lm(grades$final~grades$quiz2), col="red")

```

Scatter plot of Quiz2 vs Final



```

cor.test(grades$final, grades$quiz3)

```

```

## 
## Pearson's product-moment correlation
## 
## data: grades$final and grades$quiz3
## t = 6.8809, df = 103, p-value = 4.758e-10
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.4140461 0.6797318
## sample estimates:
## cor
## 0.5611773

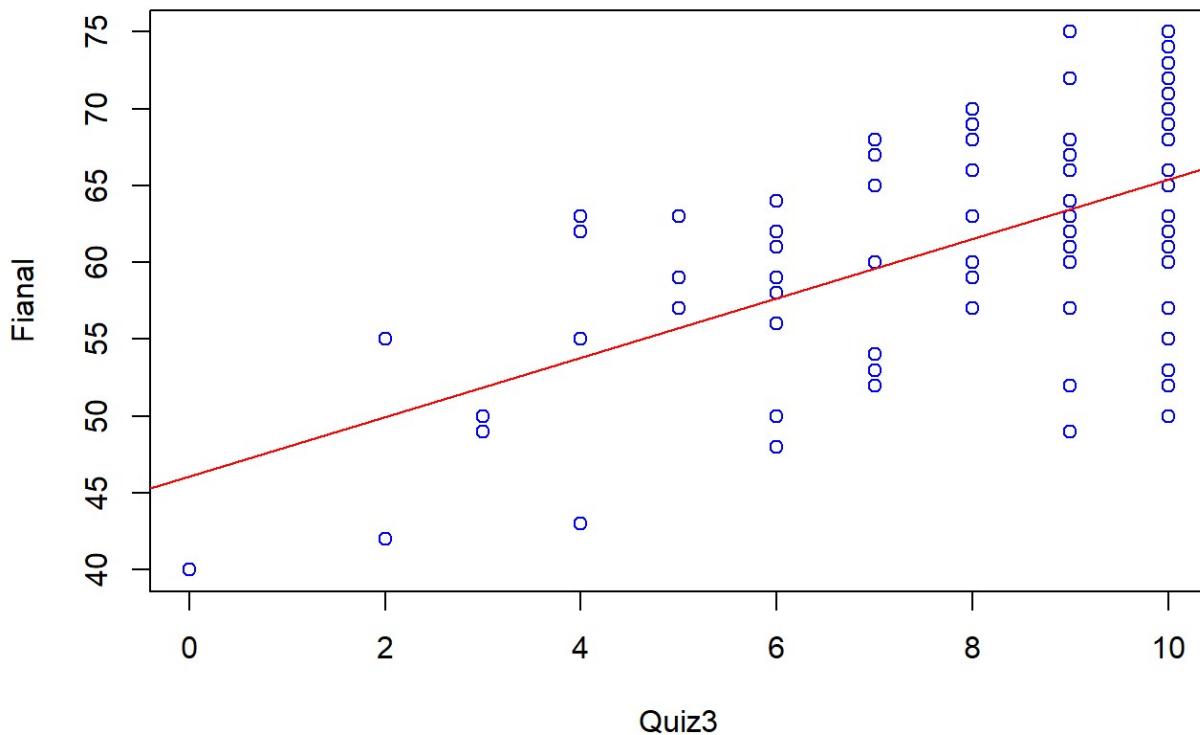
```

```

plot(final~quiz3, data = grades, main= "Scatter plot of Quiz3 vs Final", col= 'blue', y
lab= 'Final', xlab="Quiz3")
abline(lm(grades$final~grades$quiz3), col="red")

```

Scatter plot of Quiz3 vs Final



```
cor.test(grades$final, grades$quiz4)
```

```

## 
## Pearson's product-moment correlation
## 
## data: grades$final and grades$quiz4
## t = 5.6716, df = 103, p-value = 1.303e-07
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.3267180 0.6213986
## sample estimates:
## cor
## 0.4878348

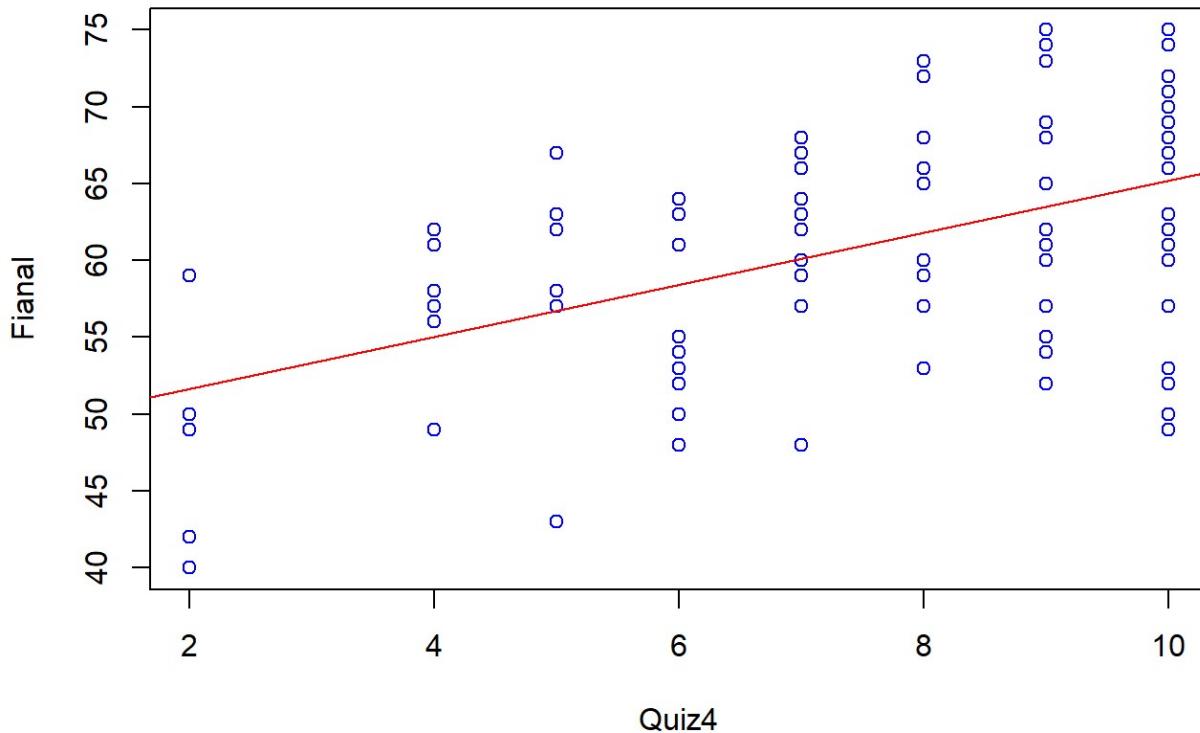
```

```

plot(final~quiz4, data = grades, main= "Scatter plot of Quiz4 vs Final", col= 'blue', y
lab= 'Final', xlab="Quiz4")
abline(lm(grades$final~grades$quiz4), col="red")

```

Scatter plot of Quiz4 vs Final



```
cor.test(grades$final, grades$quiz5)
```

```

## 
## Pearson's product-moment correlation
## 
## data: grades$final and grades$quiz5
## t = 5.4264, df = 103, p-value = 3.834e-07
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.3076485 0.6082107
## sample estimates:
## cor
## 0.4715109

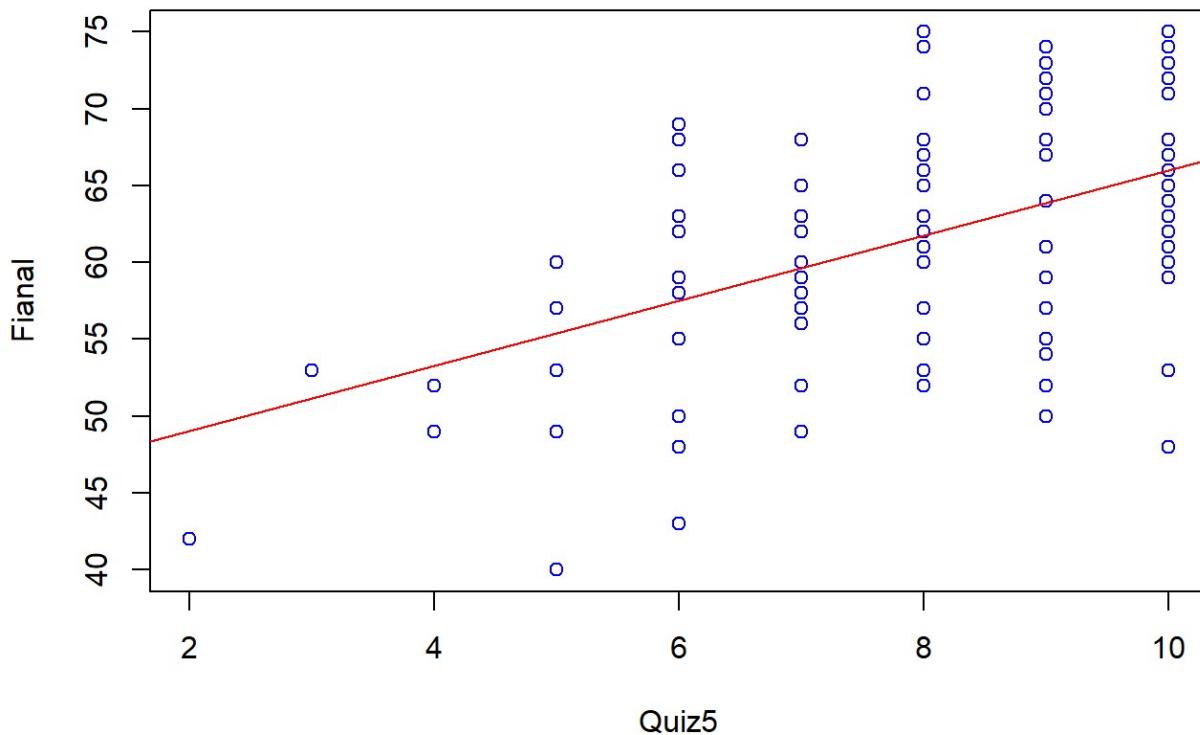
```

```

plot(final~quiz5, data = grades, main= "Scatter plot of Quiz5 vs Final", col= 'blue', y
lab= 'Final', xlab="Quiz5")
abline(lm(grades$final~grades$quiz5), col="red")

```

Scatter plot of Quiz5 vs Final



```

cor.test(grades$final, grades$total)

```

```

## 
## Pearson's product-moment correlation
## 
## data: grades$final and grades$total
## t = 19.054, df = 103, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.8316264 0.9188386
## sample estimates:
## 
## cor
## 0.8826091

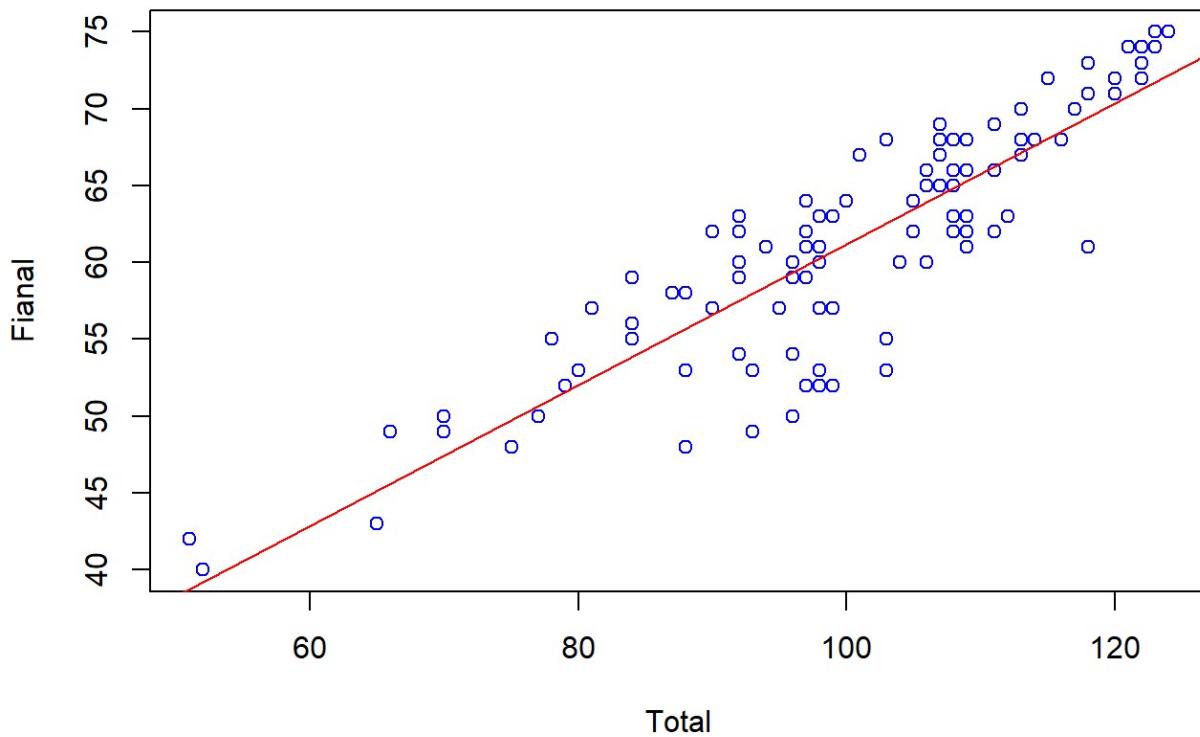
```

```

plot(final~total, data = grades, main= "Scatter plot of Total vs Final", col= 'blue', y
lab= 'Final', xlab="Total")
abline(lm(grades$final~grades$total), col="red")

```

Scatter plot of Total vs Final



```

#After correlation test start with buliding regression models and validate
model1<- lm(final ~ gpa+quiz1+quiz2+quiz3+quiz4+quiz5+total, data=grades)
model1

```

```

## 
## Call:
## lm(formula = final ~ gpa + quiz1 + quiz2 + quiz3 + quiz4 + quiz5 +
##      total, data = grades)
##
## Coefficients:
## (Intercept)          gpa        quiz1        quiz2        quiz3
## 2.2504        0.2851     -0.9128     -0.8394     -0.8627
## quiz4        quiz5        total
## -0.9338     -0.8095      0.9196

```

```
summary(model1)
```

```

## 
## Call:
## lm(formula = final ~ gpa + quiz1 + quiz2 + quiz3 + quiz4 + quiz5 +
##      total, data = grades)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -9.2806 -0.2002  0.0674  0.3430  8.1334
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 2.25044   1.05481   2.133   0.0354 *  
## gpa         0.28514   0.19778   1.442   0.1526    
## quiz1       -0.91284   0.12206  -7.478 3.38e-11 *** 
## quiz2       -0.83941   0.14401  -5.829 7.30e-08 *** 
## quiz3       -0.86271   0.12707  -6.789 9.03e-10 *** 
## quiz4       -0.93377   0.11052  -8.449 2.95e-13 *** 
## quiz5       -0.80947   0.10964  -7.383 5.37e-11 *** 
## total        0.91960   0.02192  41.943 < 2e-16 *** 
## ---        
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 
## Residual standard error: 1.34 on 97 degrees of freedom
## Multiple R-squared:  0.9735, Adjusted R-squared:  0.9715 
## F-statistic: 508.3 on 7 and 97 DF,  p-value: < 2.2e-16

```

#Lets check variance inflation factor and durbin watson statistics to check for inflation of r-square and auto correlation.

```
vif(model1)
```

```

##      gpa     quiz1     quiz2     quiz3     quiz4     quiz5     total
## 1.322065 5.312639 3.164871 4.982488 3.679254 2.170517 6.518202

```

```
durbinWatsonTest(model1)
```

```
##   lag Autocorrelation D-W Statistic p-value
##   1      -0.02699258     2.050631    0.802
## Alternative hypothesis: rho != 0
```

#In Model1, variable GPA does not have significant slope with response variable, variance inflation factor results not allowing to build a model with total as one of the predictor. Lets build one more model by removing total alone.

```
model2<- lm(final ~ gpa+quiz1+quiz2+quiz3+quiz4+quiz5, data=grades)
model2
```

```
##
## Call:
## lm(formula = final ~ gpa + quiz1 + quiz2 + quiz3 + quiz4 + quiz5,
##      data = grades)
##
## Coefficients:
## (Intercept)          gpa        quiz1        quiz2        quiz3
##            32.8658      3.6271      0.2986      0.8004      0.9063
##      quiz4        quiz5
##            -0.1428      0.4823
```

```
summary(model2)
```

```

## 
## Call:
## lm(formula = final ~ gpa + quiz1 + quiz2 + quiz3 + quiz4 + quiz5,
##      data = grades)
##
## Residuals:
##    Min     1Q Median     3Q    Max
## -13.6580 -2.7187  0.7985  3.9664 11.4979
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 32.8658   3.3141   9.917 < 2e-16 ***
## gpa         3.6271   0.7879   4.604 1.25e-05 ***
## quiz1       0.2986   0.5162   0.578  0.5643    
## quiz2       0.8004   0.6032   1.327  0.1876    
## quiz3       0.9063   0.5217   1.737  0.0855 .  
## quiz4      -0.1428   0.4739  -0.301  0.7638    
## quiz5       0.4823   0.4580   1.053  0.2948    
## ---      
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.831 on 98 degrees of freedom
## Multiple R-squared:  0.4922, Adjusted R-squared:  0.4611 
## F-statistic: 15.83 on 6 and 98 DF,  p-value: 1.224e-12

```

```
vif(model2)
```

```

##      gpa    quiz1    quiz2    quiz3    quiz4    quiz5
## 1.107509 5.015198 2.931613 4.433567 3.572136 1.999252

```

```
durbinWatsonTest(model2)
```

```

## lag Autocorrelation D-W Statistic p-value
## 1    -0.1175758    2.224256   0.244
## Alternative hypothesis: rho != 0

```

#From Model2, VIF result it concluded that quiz1 is increasing corelation artificially, hence it will be removed from model an new model is constructed as follows

```

model3<- lm(final ~ gpa+quiz2+quiz3+quiz4+quiz5, data=grades)
model3

```

```
##  
## Call:  
## lm(formula = final ~ gpa + quiz2 + quiz3 + quiz4 + quiz5, data = grades)  
##  
## Coefficients:  
## (Intercept)      gpa      quiz2      quiz3      quiz4  
## 32.49410     3.62466     0.82825     1.06375    -0.02019  
##      quiz5  
## 0.50431
```

```
summary(model3)
```

```
##  
## Call:  
## lm(formula = final ~ gpa + quiz2 + quiz3 + quiz4 + quiz5, data = grades)  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max  
## -13.5494  -2.7210   0.5787   4.0927  11.5517  
##  
## Coefficients:  
##             Estimate Std. Error t value Pr(>|t|)  
## (Intercept) 32.49410   3.24023 10.028 < 2e-16 ***  
## gpa         3.62466   0.78519  4.616 1.17e-05 ***  
## quiz2       0.82825   0.59928  1.382  0.1701  
## quiz3       1.06375   0.44360  2.398  0.0184 *  
## quiz4      -0.02019   0.42244 -0.048  0.9620  
## quiz5       0.50431   0.45486  1.109  0.2702  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 5.812 on 99 degrees of freedom  
## Multiple R-squared:  0.4904, Adjusted R-squared:  0.4647  
## F-statistic: 19.06 on 5 and 99 DF,  p-value: 3.089e-13
```

```
vif(model3)
```

```
##      gpa      quiz2      quiz3      quiz4      quiz5  
## 1.107478 2.912966 3.227419 2.857258 1.985498
```

```
durbinWatsonTest(model3)
```

```
## lag Autocorrelation D-W Statistic p-value
##    1      -0.1288976    2.247614   0.202
## Alternative hypothesis: rho != 0
```

```
#From model3 summary and VIF Out put, it was decided that though no inflators in model
#3.variables quiz2, quiz4 and quiz5 does not have significant pvalue they were removed
#and model as follows
model4<- lm(final ~ gpa+quiz3, data=grades)
model4
```

```
##
## Call:
## lm(formula = final ~ gpa + quiz3, data = grades)
##
## Coefficients:
## (Intercept)      gpa      quiz3
##       37.541     3.993     1.609
```

```
summary(model4)
```

```
##
## Call:
## lm(formula = final ~ gpa + quiz3, data = grades)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.5260  -3.2000  -0.0803   4.7388  12.3955
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 37.5415    2.6802 14.007 < 2e-16 ***
## gpa         3.9926    0.7850  5.086 1.67e-06 ***
## quiz3       1.6088    0.2598  6.193 1.26e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.929 on 102 degrees of freedom
## Multiple R-squared:  0.4535, Adjusted R-squared:  0.4428
## F-statistic: 42.32 on 2 and 102 DF,  p-value: 4.133e-14
```

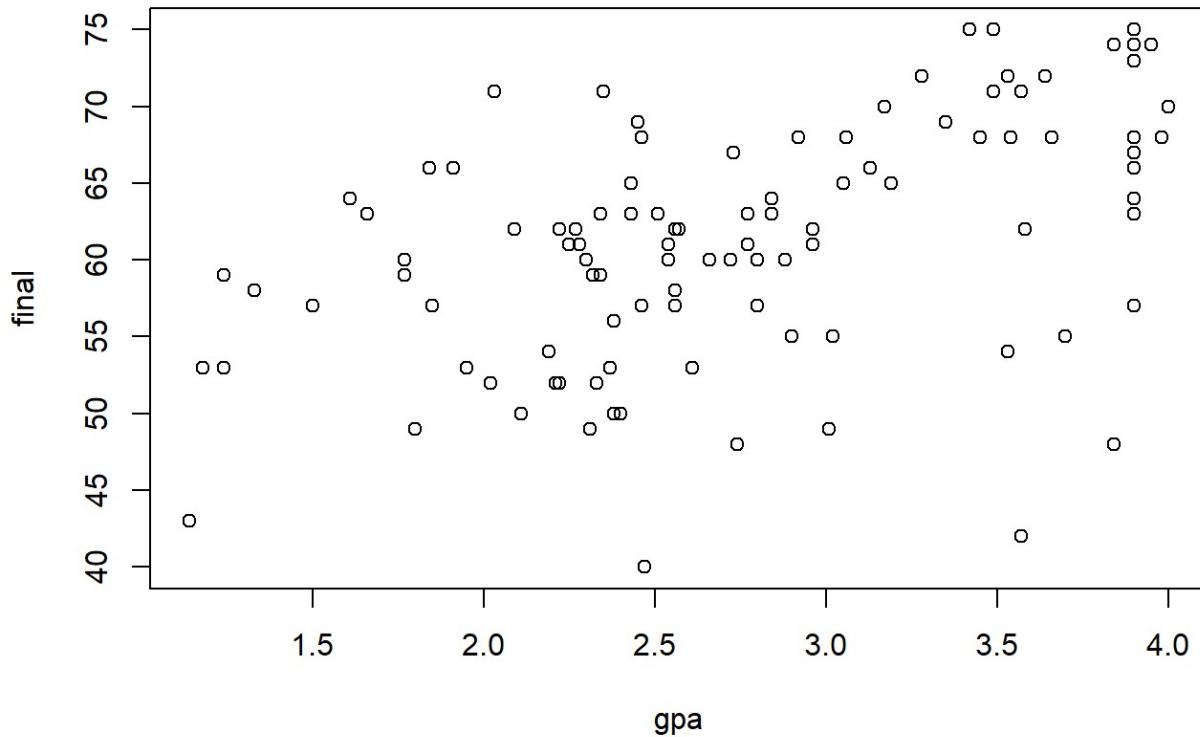
```
vif(model4)
```

```
##      gpa    quiz3
## 1.06341 1.06341
```

```
durbinWatsonTest(model4)
```

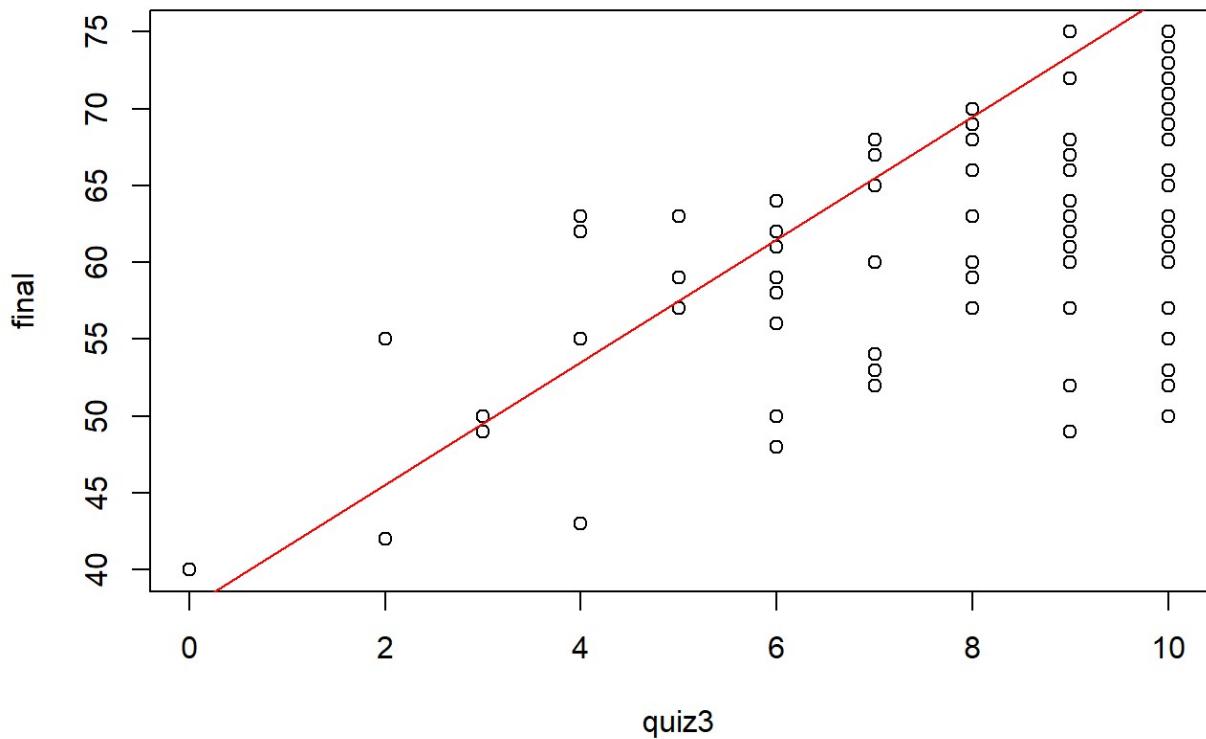
```
##   lag Autocorrelation D-W Statistic p-value
##   1    -0.09929107     2.194562    0.31
## Alternative hypothesis: rho != 0
```

```
#from model4 out put and vif, both predictors can be used for regression, lets plot final vs predictors of GPA and quiz3
plot(final~gpa+quiz3, data = grades)
```

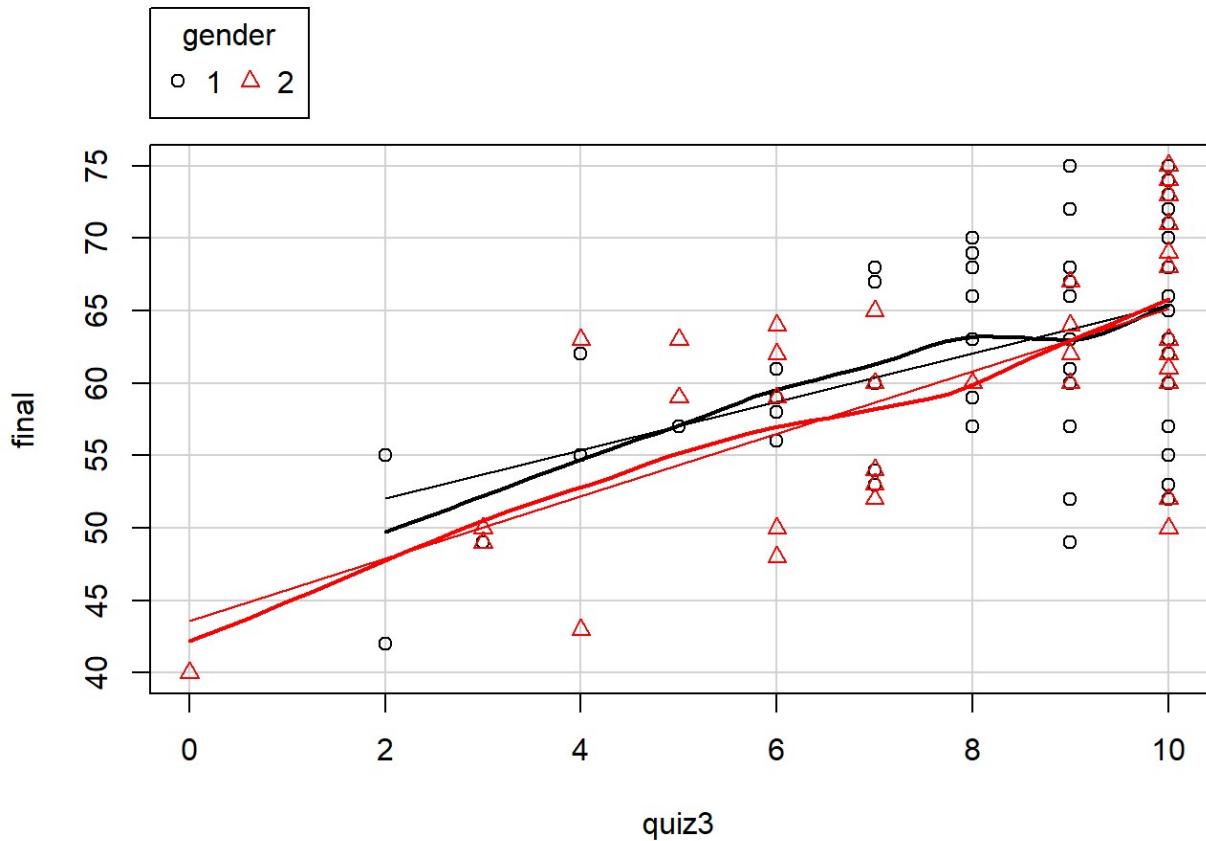


```
abline(lm(grades$final~grades$gpa+grades$quiz3), col="red")
```

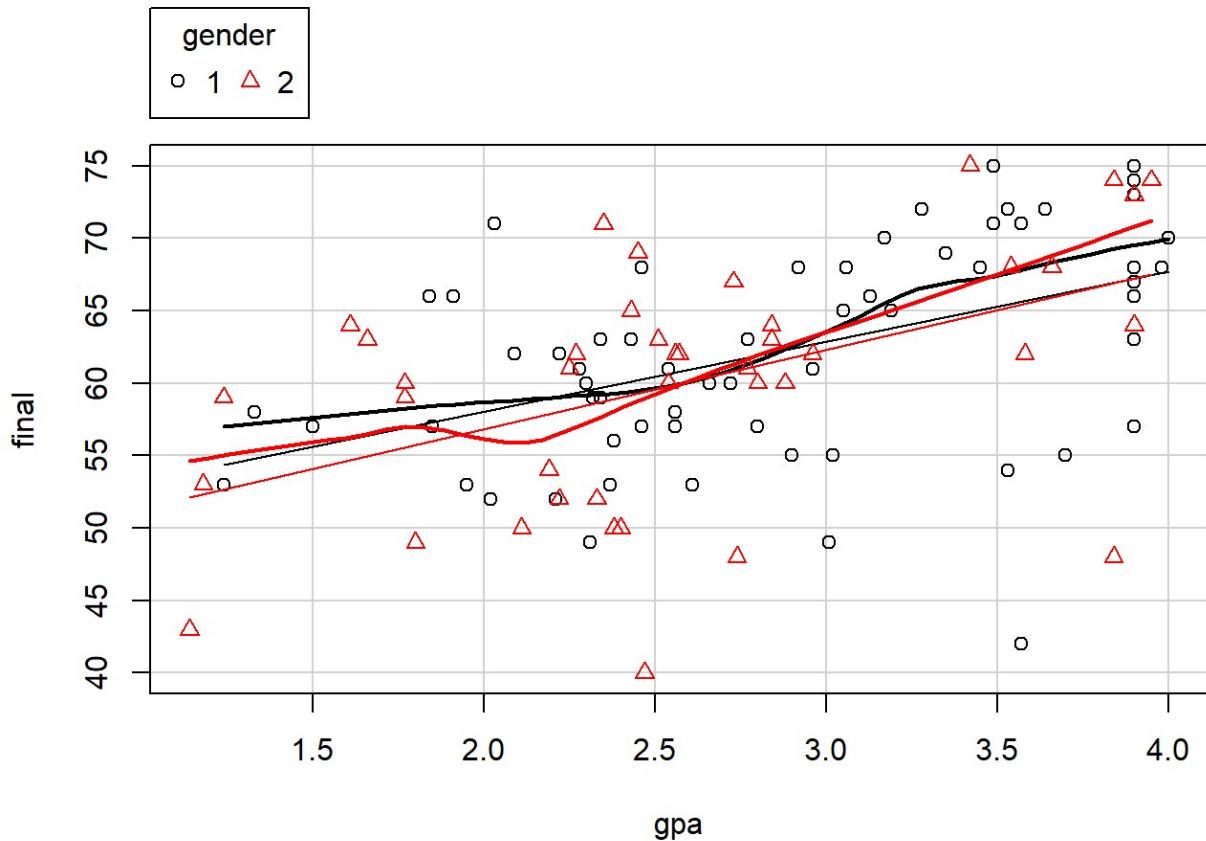
```
## Warning in abline(lm(grades$final ~ grades$gpa + grades$quiz3), col =
## "red"): only using the first two of 3 regression coefficients
```



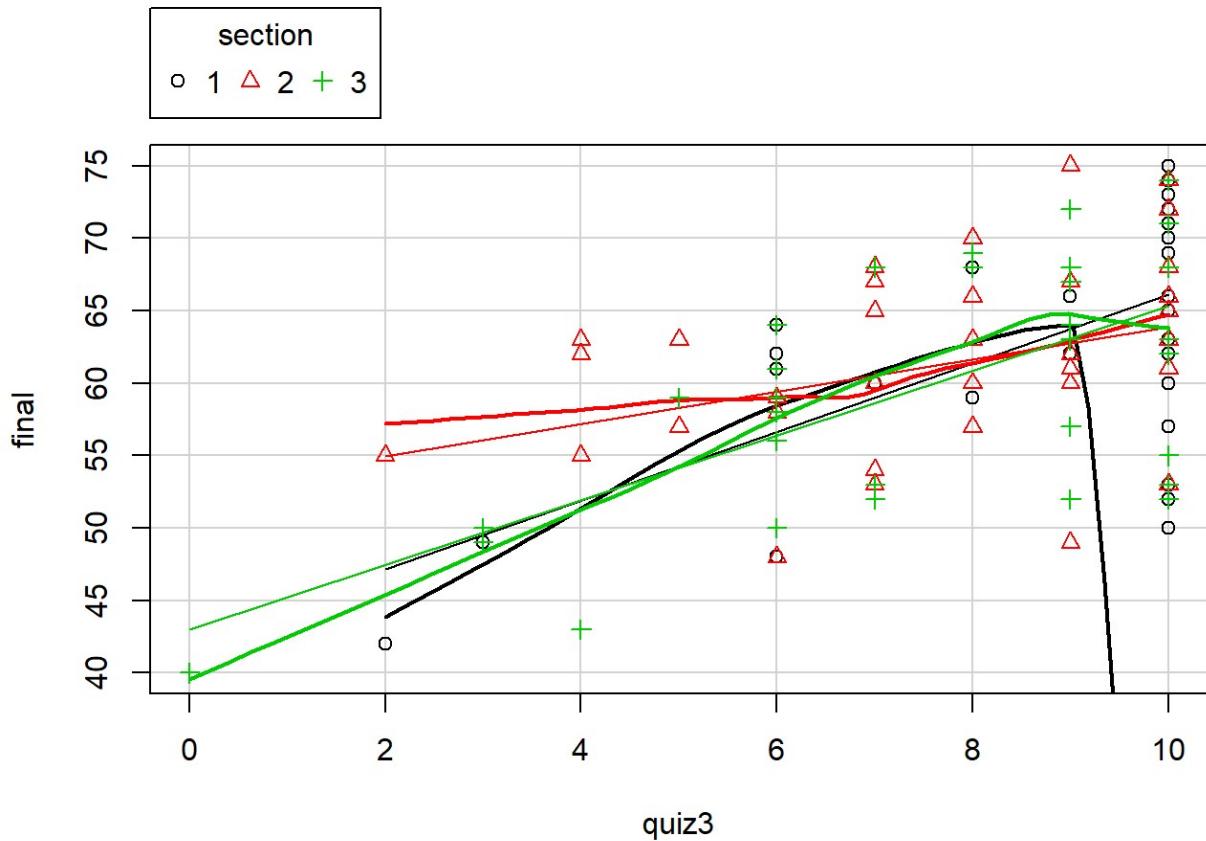
```
scatterplot(final~quiz3|gender, data=grades)
```



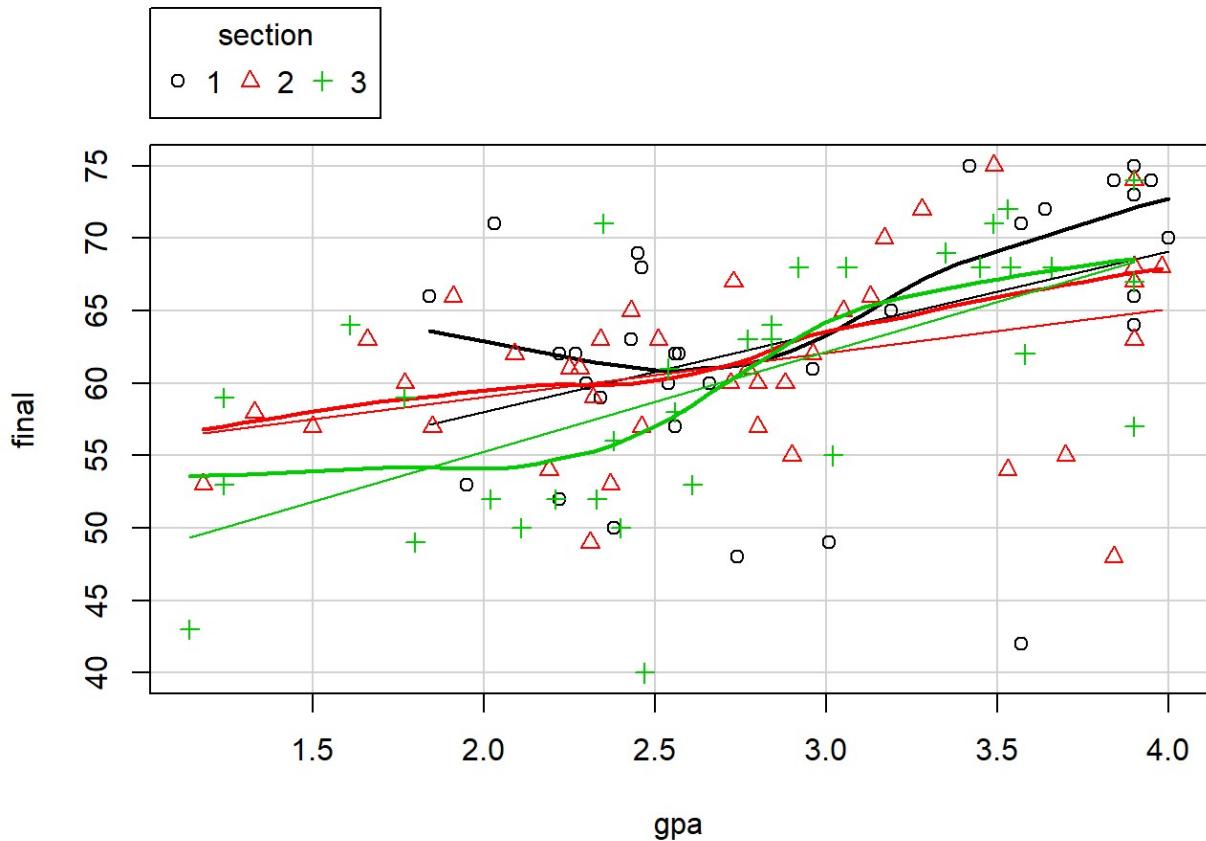
```
scatterplot(final~gpa|gender, data=grades)
```



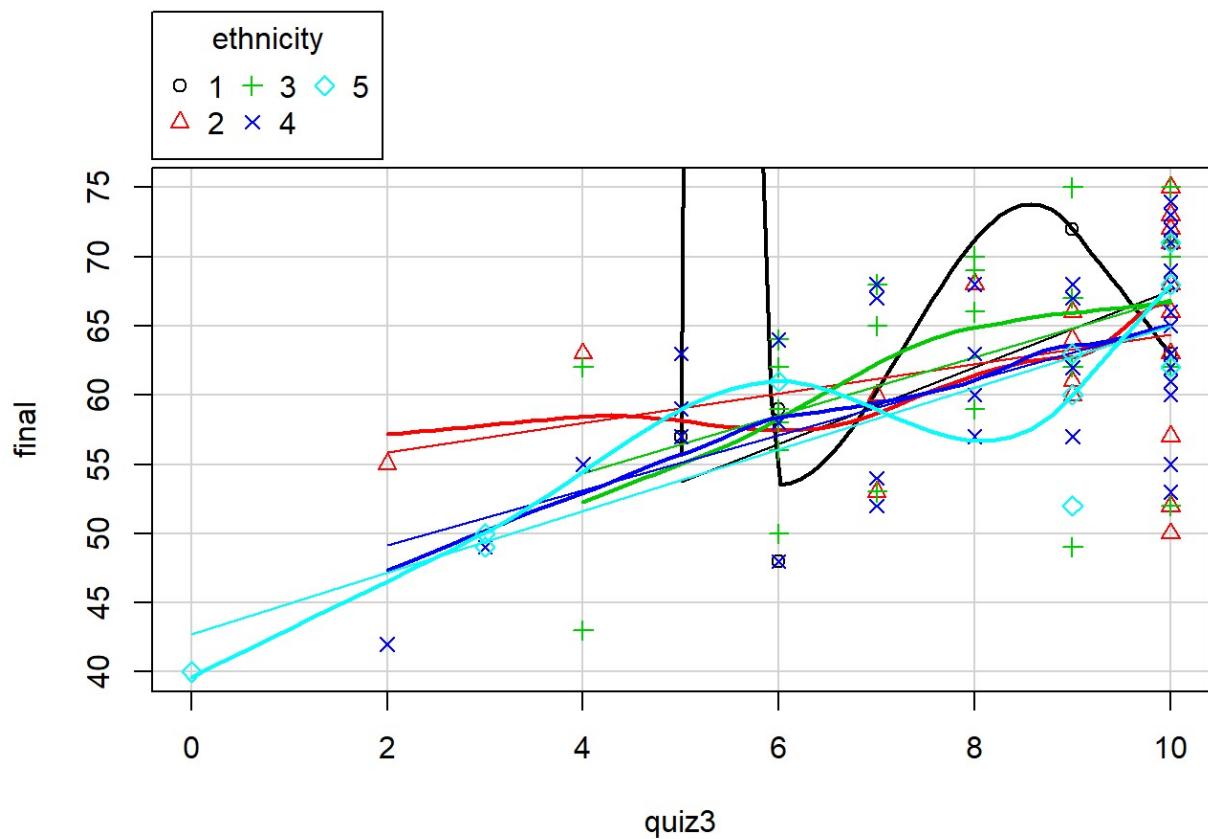
```
scatterplot(final~quiz3|section, data=grades)
```



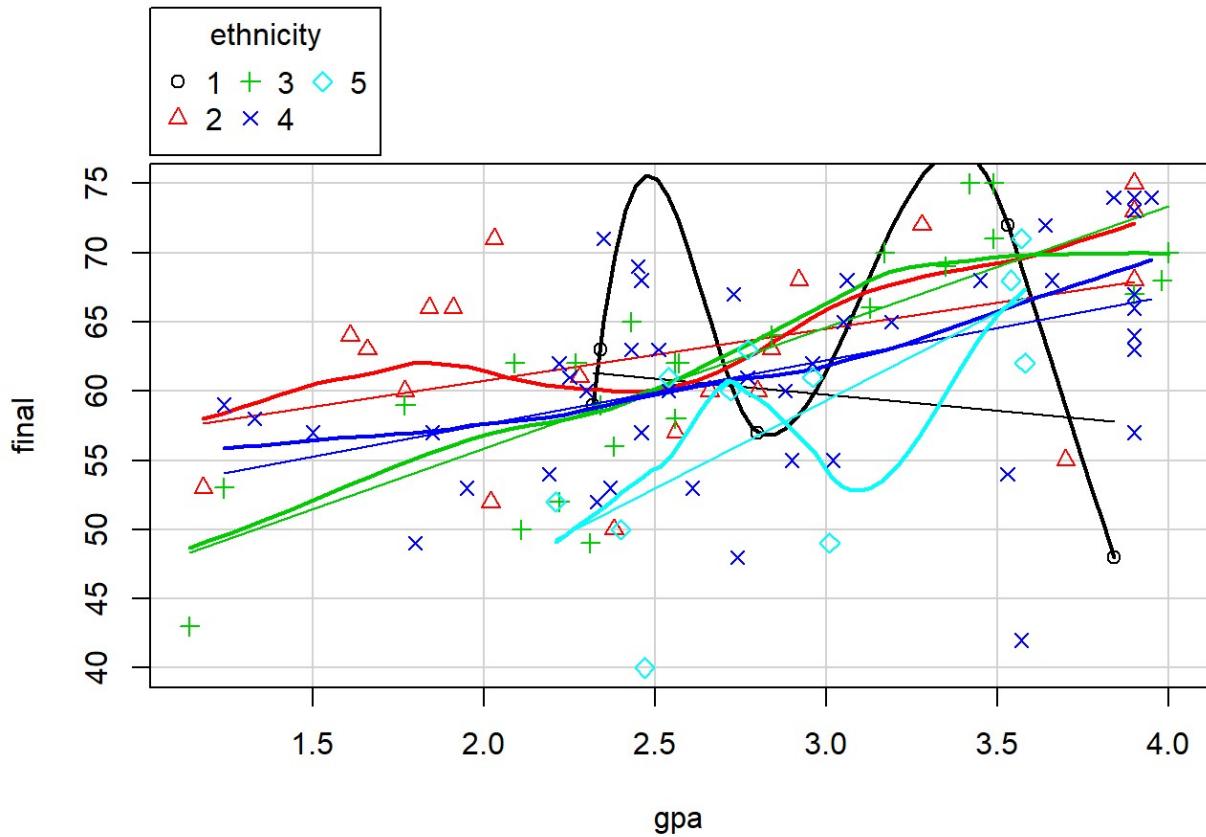
```
scatterplot(final~gpa|section, data=grades)
```



```
scatterplot(final~quiz3|ethnicity, data=grades)
```

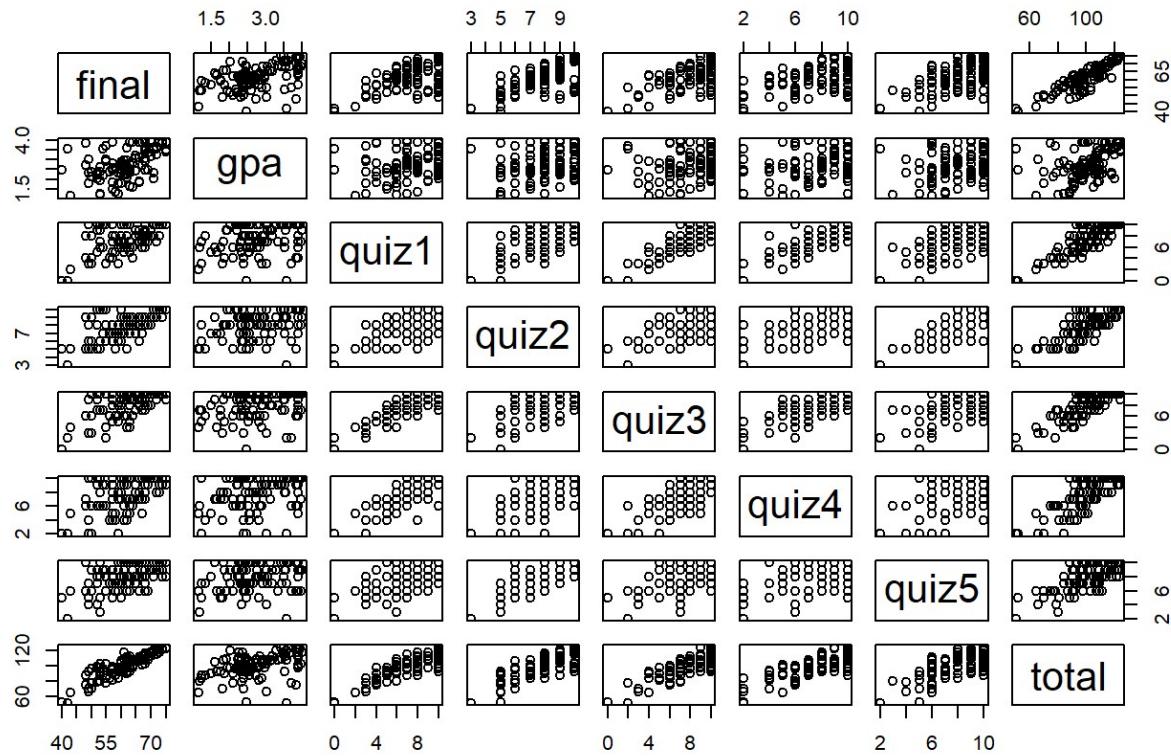


```
scatterplot(final~gpa|ethnicity, data=grades)
```



```
#visualize association of predictors and response variable.
pairs(~final+gpa+quiz1+quiz2+quiz3+quiz4+quiz5+total, data=grades, main= "scatter plot matrix")
```

scatter plot matrix



#Model4 is having R-squared value less when compared with model3.. so, model3 will be selected as final model

#4. How do you interpret significance value of F-statistics? Mention in 4 Lines and show R Output. [Fitness of model]

```
anova(model3)
```

```
## Analysis of Variance Table
##
## Response: final
##             Df Sum Sq Mean Sq F value    Pr(>F)
## gpa         1 1627.8 1627.81 48.1944 4.077e-10 ***
## quiz2       1 1226.1 1226.12 36.3016 2.898e-08 ***
## quiz3       1   322.7   322.75  9.5555  0.002589 **
## quiz4       1     0.2     0.18  0.0054  0.941762
## quiz5       1    41.5    41.52  1.2293  0.270235
## Residuals  99 3343.8   33.78
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
anova(model4)
```

```
## Analysis of Variance Table
##
## Response: final
##           Df Sum Sq Mean Sq F value    Pr(>F)
## gpa        1 1627.8 1627.81 46.300 7.096e-10 ***
## quiz3      1 1348.3 1348.28 38.349 1.255e-08 ***
## Residuals 102 3586.1   35.16
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#From anova test p value it can conclude that quiz2, quiz4 and quiz5 are not having significant slope with response variable fianl So, lets Check durbin watson statistic test to check auto corrrlation among variables.

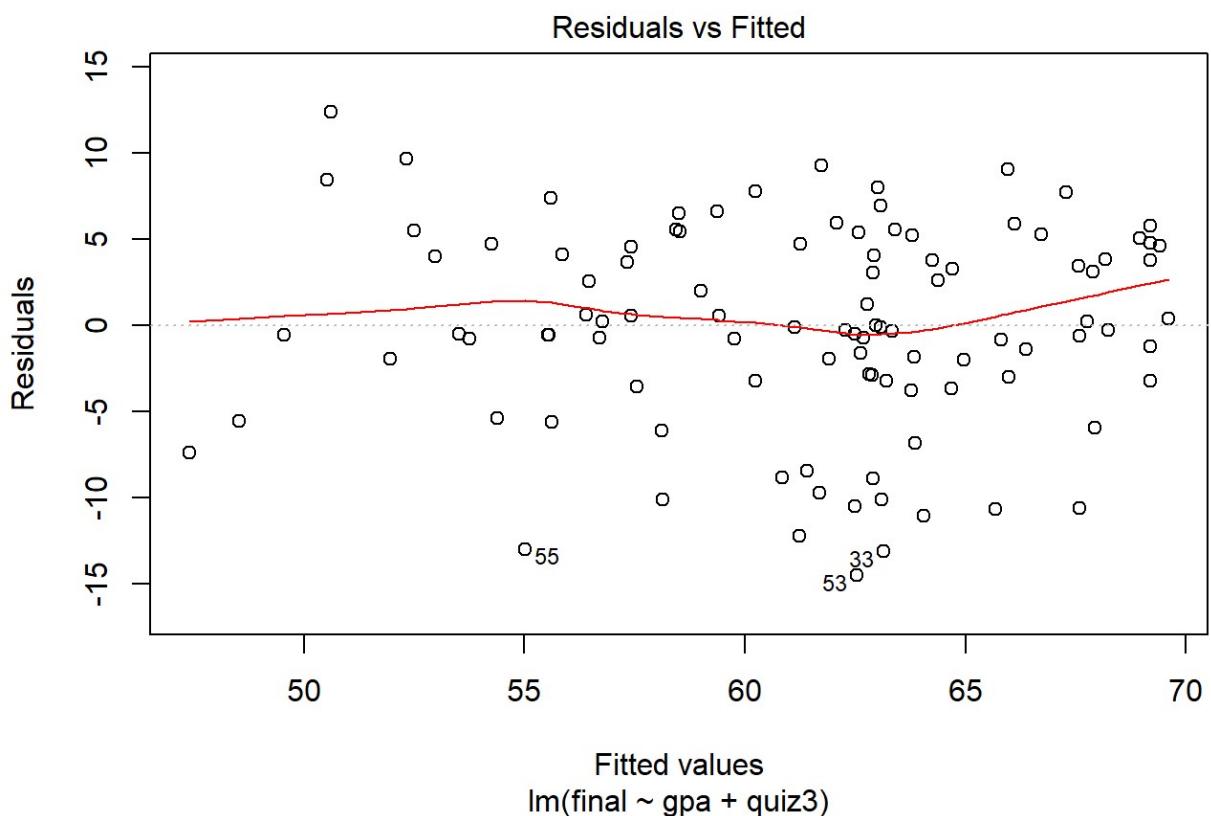
```
durbinWatsonTest(model3)
```

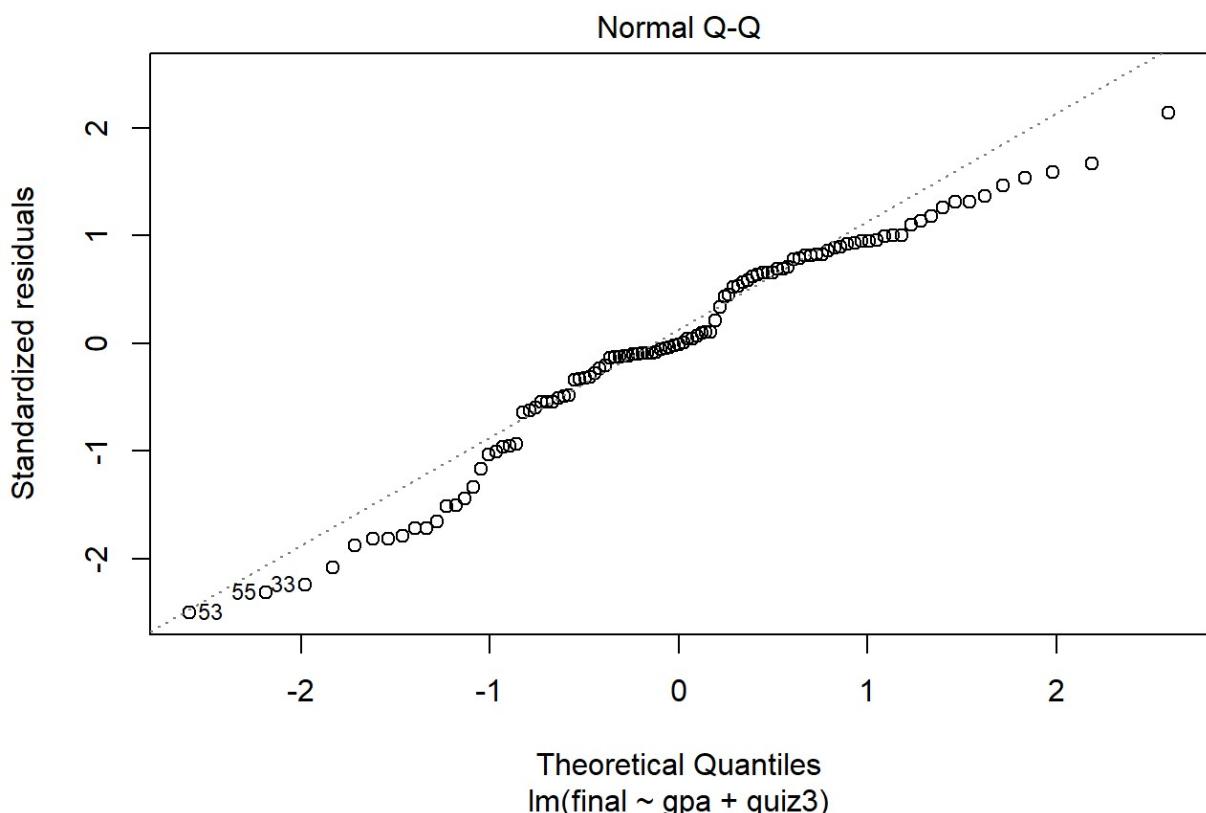
```
## lag Autocorrelation D-W Statistic p-value
## 1     -0.1288976    2.247614   0.232
## Alternative hypothesis: rho != 0
```

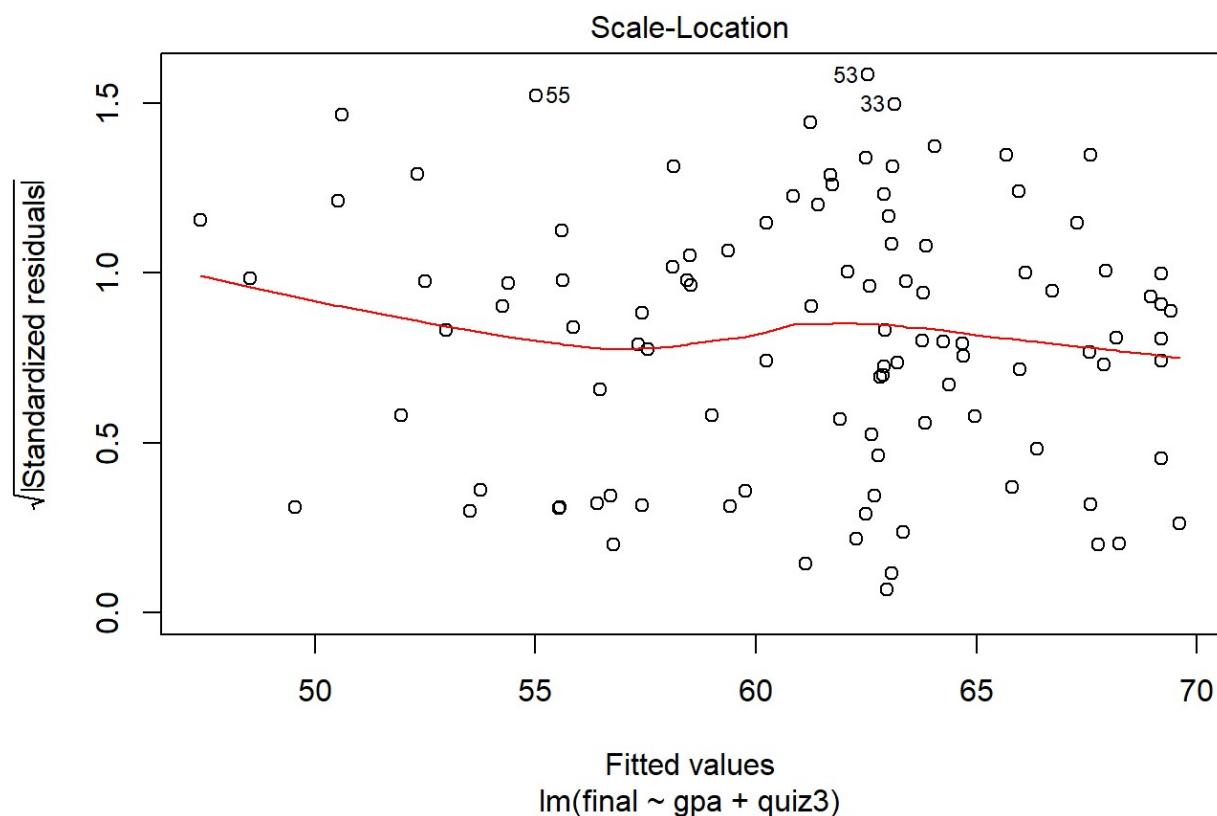
```
durbinWatsonTest(model4)
```

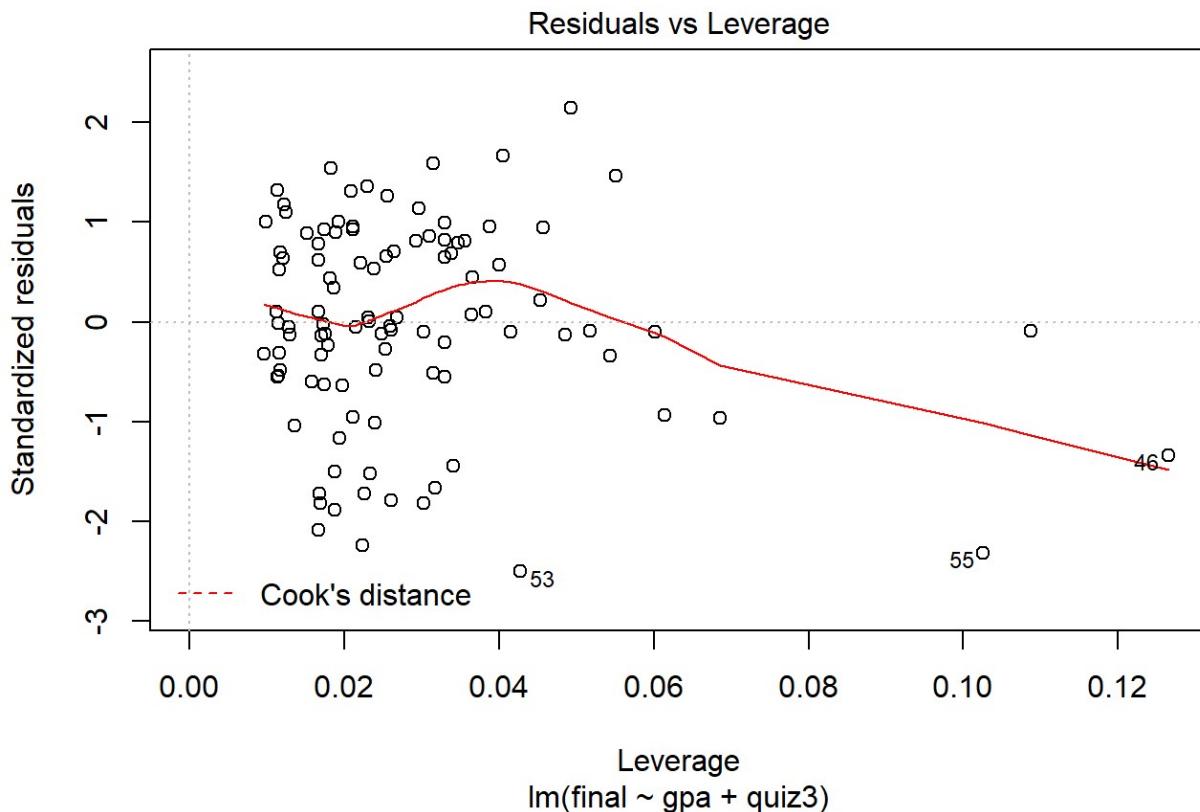
```
## lag Autocorrelation D-W Statistic p-value
## 1     -0.09929107   2.194562   0.35
## Alternative hypothesis: rho != 0
```

```
plot(model4)
```









#9. Test the assumption of Normality and interpret your findings. [hint: histogram of residuals/errors]

#Show histogram and interpret in maximum 3 lines.

#Lets find error/residual term of model3 as below

```
residual<-residuals(model4)
residual
```

```

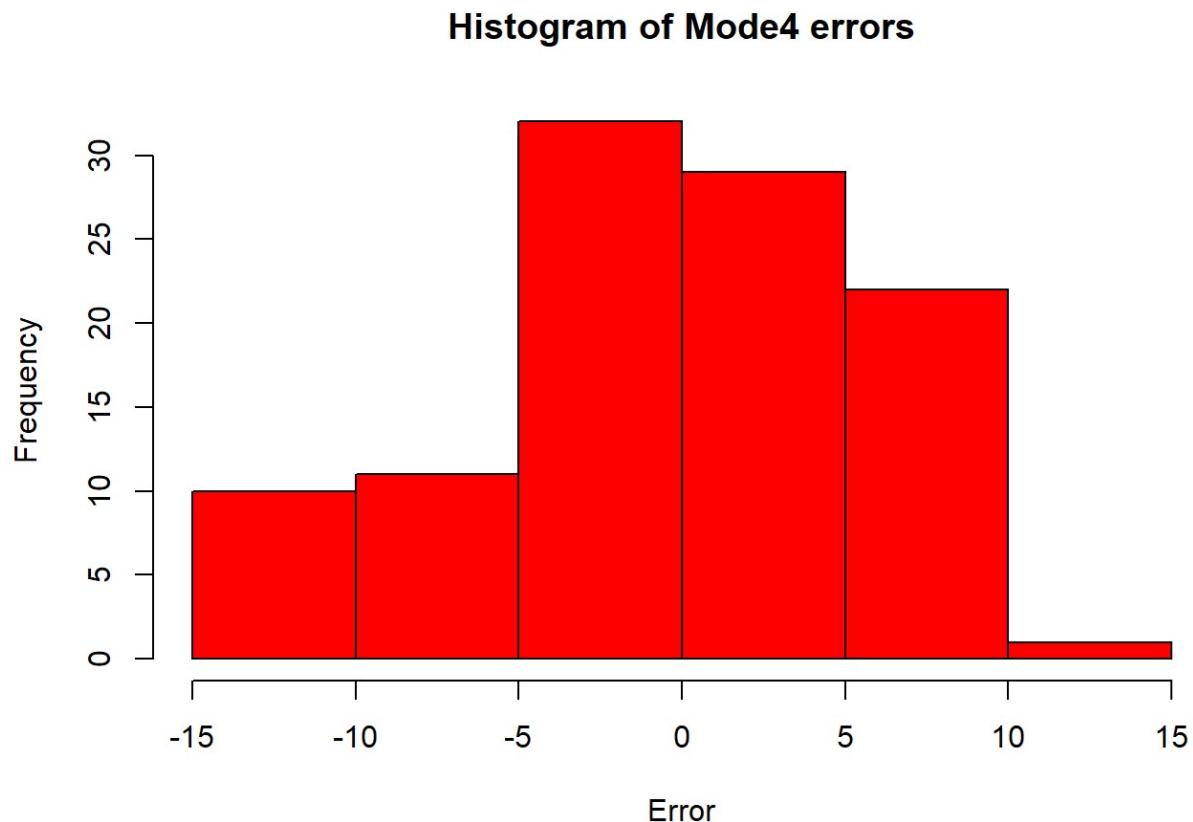
##      1          2          3          4          5
## -0.51437366 -3.54694323 -3.23375140  3.30622554  6.63289284
##      6          7          8          9         10
##  4.79925345 -1.96854482  3.11682569  4.59962129  9.04503165
##     11         12         13         14         15
##  2.54280767  7.98785032  5.58858601 -0.55533632 -6.10591327
##     16         17         18         19         20
##  3.66442618 12.39554117 -10.13410245  7.39299999 -3.77075188
##     21         22         23         24         25
## -0.53186180 -1.20074655 -8.84438513  7.76624860 -5.52828439
##     26         27         28         29         30
##  9.67870462  0.02777675  9.26549612  0.23513348 -1.61288537
##     31         32         33         34         35
## -0.12387015 -10.68722058 -13.13192897  4.74461330 -8.89708504
##     36         37         38         39         40
## -11.05023690 -3.20004458  0.57651449 -0.08026529  5.03881204
##     41         42         43         44         45
## -0.69675092 -0.75463422  6.93147197  7.71572216 -2.88063313
##     46         47         48         49         50
## -7.40332171  5.49552438 -2.98315752  4.79925345  1.98751606
##     51         52         53         54         55
##  5.79925345  5.21279621 -14.52600990  5.42232641 -13.01281820
##     56         57         58         59         60
## -0.80699988 -0.75393225  4.73876140  3.83734067  0.58457332
##     61         62         63         64         65
##  8.46365678  0.59918602 -0.59195203 -5.92310074  5.55120076
##     66         67         68         69         70
##  4.12996688 -0.28173666 -1.83886748  3.09117770 -10.59195203
##     71         72         73         74         75
## -0.69273823  6.49482241 -1.91066152 -1.95022024 -5.38573255
##     76         77         78         79         80
##  3.79925345  0.39998914 -0.49310607  4.58457332 -8.41509242
##     81         82         83         84         85
## -3.20074655  0.23660498 -0.24251219 -6.85060474  5.92963276
##     86         87         88         89         90
##  3.79925345 -12.24364944 -3.68905980  5.46663324  4.02814446
##     91         92         93         94         95
##  3.43623714 -10.49310607  5.27469220  4.07944044 -9.69457744
##     96         97         98         99        100
##  5.88532593  2.62563700 -2.81251752 -10.09200254 -1.36596992
##    101        102        103        104        105
##  1.23443151 -0.55463435 -5.61873727 -0.33156113  3.76186820

```

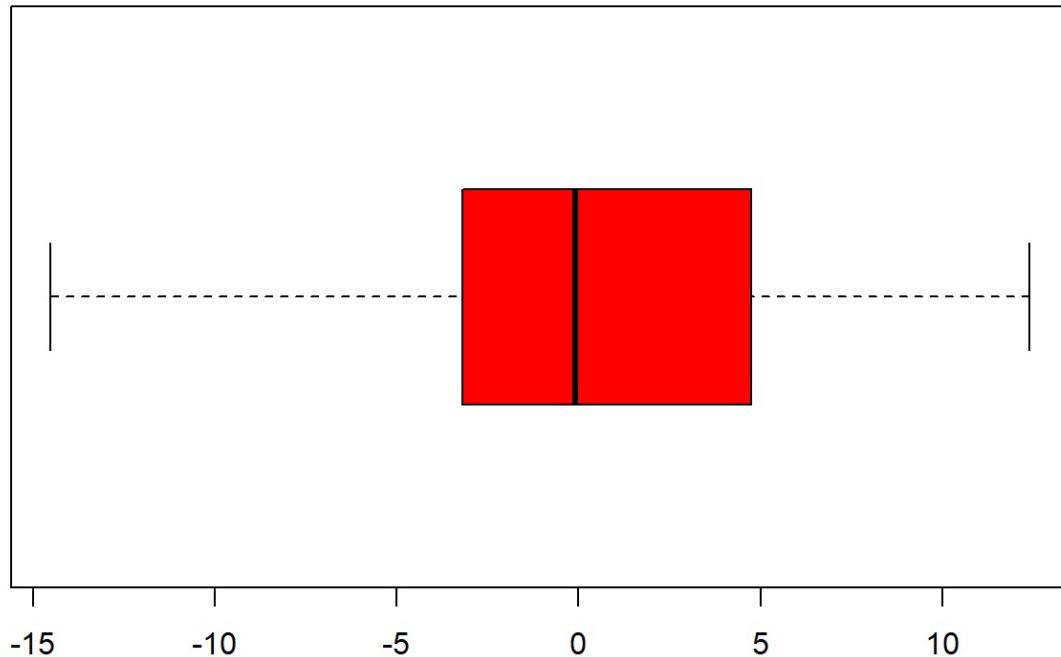
```
summary(residual)
```

```
##      Min.    1st Qu.     Median      Mean    3rd Qu.      Max.  
## -14.52601 -3.20004 -0.08027  0.00000  4.73876 12.39554
```

```
hist(residual, col="red", main = "Histogram of Mode4 errors", xlab = "Error", ylab =  
"Frequency")
```



```
boxplot(residual, col="red", horizontal = T)
```



#10. Test the assumption of Independent of observations and interpret in maximum 3 lines [hint: draw scatter plot between residuals/errors (y-axis) and observation numbers (x-axis)]

```
grades$residual=NULL  
grades$residual=residual  
obs_no<-c(1:105)  
grades$obs_no=NULL  
grades$obs_no=obs_no  
grades
```

##	Sr_No	id	lastname	firstname	gender	ethnicity	year	lowup	section
## 1	1	106484	VILLARRUZ	ALFRED	2		2	2	1
## 2	2	108642	VALAZQUEZ	SCOTT	2		4	3	2
## 3	3	127285	GALVEZ	JACKIE	1		4	4	2
## 4	4	132931	OSBORNE	ANN	1		3	2	1
## 5	5	140219	GUADIZ	VALERIE	1		2	4	2
## 6	6	142630	RANGIFO	TANIECE	1		4	3	2
## 7	7	153964	TOMOSAWA	DANIEL	2		2	3	2
## 8	8	154441	LIAN	JENNY	1		5	2	1
## 9	9	157147	BAKKEN	KREG	2		4	3	2
## 10	10	164605	LANGFORD	DAWN	1		3	3	2
## 11	11	164842	VALENZUELA	NANCY	1		1	4	2
## 12	12	167664	SWARM	MARK	2		4	3	2
## 13	13	175325	KHOURY	DENNIS	2		4	3	2
## 14	14	192627	MISCHKE	ELAINE	1		4	1	1
## 15	15	211239	AUSTIN	DERRICK	2		4	3	2
## 16	16	219593	POTTER	MICKEY	1		5	3	2
## 17	17	237983	LEE	JONATHAN	2		2	4	2
## 18	18	245473	DAYES	ROBERT	2		4	3	2
## 19	19	249586	STOLL	GLENDON	2		4	3	2
## 20	20	260983	CUSTER	JAMES	2		4	4	2
## 21	21	273611	WU	VIDYUTH	1		2	2	1
## 22	22	280440	CHANG	RENE	1		2	3	2
## 23	23	287617	CUMMINGS	DAVENA	1		5	3	2
## 24	24	289652	BRADLEY	SHANNON	1		4	3	2
## 25	25	302400	JONES	ROBERT	2		3	4	2
## 26	26	307894	TORRENCE	GWEN	1		3	2	1
## 27	27	337908	UYEYAMA	VICTORINE	1		1	3	2
## 28	28	354601	CARPIO	MARY	1		2	2	1
## 29	29	378446	SAUNDERS	TAMARA	1		1	2	1
## 30	30	380157	LUTZ	WILLIAM	2		4	3	2
## 31	31	390203	SHIMA	MIHAELA	1		2	3	2
## 32	32	392464	DOMINGO	MONIKA	1		4	3	2
## 33	33	414775	RATANA	JASON	2		2	3	2
## 34	34	417003	EVANGELIST	NIKKI	1		2	3	2
## 35	35	419891	DE CANIO	PAULA	1		4	3	2
## 36	36	420327	BADGER	SUZANNA	1		4	3	2
## 37	37	434571	SURI	MATHEW	2		2	3	2
## 38	38	436413	PANG	SUZANNE	1		2	3	2
## 39	39	447659	GALANVILLE	DANA	1		5	4	2
## 40	40	463276	HANSEN	TIM	2		4	3	2
## 41	41	466407	PICKERING	HEIDI	1		3	3	2
## 42	42	467806	DEVERS	GAIL	1		3	3	2
## 43	43	473303	PARK	SANDRA	1		3	4	2
## 44	44	479547	LANGFORD	BLAIR	2		3	3	2
## 45	45	490016	STEPHEN	LIZA	1		5	3	2
## 46	46	498900	HUANG	<td>2</td> <td></td> <td>5</td> <td>3</td> <td>2</td>	2		5	3	2
## 47	47	506467	SCARBROUGH	CYNTHE	1		4	3	2

## 48	48	515586	FIALLOS	LAUREL	1	4	2	1	2
## 49	49	519444	RATHBUN	DAWNE	1	4	4	2	2
## 50	50	546022	HAMIDI	KIMBERLY	1	5	3	2	1
## 51	51	553919	KWON	SHELLY	1	2	3	2	1
## 52	52	554809	JONES	LISA	1	3	3	2	3
## 53	53	574170	HURRIA	WAYNE	2	1	2	1	2
## 54	54	576008	BULMERKA	HUSIBA	1	4	4	2	3
## 55	55	576141	MISHALANY	LUCY	1	4	3	2	1
## 56	56	594463	CRUZADO	MARITESS	1	4	4	2	2
## 57	57	595177	WILLIAMS	OLIMPIA	1	3	3	2	3
## 58	58	615115	VASENIUS	RUSS	2	3	3	2	3
## 59	59	616095	SPRINGER	ANNELIES	1	4	3	2	1
## 60	60	623857	CORTEZ	VIKKI	1	3	4	2	3
## 61	61	664653	KHAN	JOHN	2	4	3	2	3
## 62	62	681855	GRISWOLD	TAMMY	1	4	3	2	2
## 63	63	700978	WEBSTER	DEANNA	1	3	2	1	3
## 64	64	703740	SUNYA	DALE	2	5	3	2	3
## 65	65	721311	SONG	LOIS	2	2	3	2	3
## 66	66	725987	BATILLER	FRED	2	2	2	1	2
## 67	67	737728	BELTRAN	JIM	2	3	3	2	1
## 68	68	755724	LANGFORD	TREVOR	2	4	3	2	2
## 69	69	756097	KURSEE	JACKIE	1	3	3	2	2
## 70	70	762308	GOUW	BONNIE	1	4	2	1	3
## 71	71	762813	DAEL	IVAN	2	3	2	1	1
## 72	72	765360	ROBINSON	ERIC	2	3	3	2	2
## 73	73	768995	DUMITRESCU	STACY	2	4	4	2	2
## 74	74	777683	ANDERSON	ERIC	2	5	4	2	3
## 75	75	779481	AHGHEL	BRENDA	1	5	3	2	1
## 76	76	780028	ROBINSON	CLAYTON	2	4	3	2	1
## 77	77	781676	WATKINS	YVONNE	1	3	4	2	1
## 78	78	798931	ZUILL	RENAE	1	4	3	2	1
## 79	79	807963	LEWIS	CARL	2	3	2	1	1
## 80	80	818528	CARRINGTON	JYLL	1	4	3	2	1
## 81	81	822485	VALENZUELA	KATHRYN	1	4	1	1	1
## 82	82	843472	PRADO	DON	2	5	3	2	3
## 83	83	870810	REYNO	NICHOLAS	2	4	3	2	3
## 84	84	896972	HUANG	MIRNA	1	2	3	2	1
## 85	85	897606	GENOBAGA	JACQUELINE	1	2	3	2	3
## 86	86	898766	RAO	DAWN	1	2	3	2	1
## 87	87	899529	HAWKINS	CARHERINE	1	3	4	2	2
## 88	88	900485	COCHRAN	STACY	2	4	3	2	2
## 89	89	905109	JENKINS	ERIC	2	3	2	1	3
## 90	90	908754	MARQUEZ	CHYRELLE	1	4	1	1	2
## 91	91	911355	LESKO	LETICIA	1	3	2	1	3
## 92	92	915457	SHEARER	LUCIO	2	3	3	2	1
## 93	93	920656	LIAO	MICHELLE	1	2	2	1	2
## 94	94	921297	KINZER	RICHARD	2	4	3	2	2
## 95	95	938666	SUAREZ-TAN	KHANH	1	2	3	2	3
## 96	96	938881	YEO	DENISE	1	1	3	2	3

## 97	97	944702	LEDESMA	MARTINE	1	4	3	2	2		
## 98	98	958384	RONCO	SHERRY	1	4	2	1	1		
## 99	99	972678	KAHRS	JANN	1	4	4	2	2		
## 100	100	973427	ROSS	MARIA	1	4	4	2	1		
## 101	101	978889	ZIMCHEK	ARMANDO	2	4	4	2	1		
## 102	102	979028	NEUHARTH	JIM	2	4	3	2	3		
## 103	103	983522	SLOAT	AARON	2	3	3	2	3		
## 104	104	985700	CHA	LILY	1	4	2	1	1		
## 105	105	988808	MCCONAHA	CORA	1	4	3	2	3		
## gpa extrc review quiz1 quiz2 quiz3 quiz4 quiz5 final total percent											
## 1	1.18	1	2	6	5	7	6	3	53	80	64
## 2	2.19	2	1	10	10	7	6	9	54	96	77
## 3	2.46	2	2	10	7	8	9	7	57	98	78
## 4	3.98	1	1	7	8	7	7	6	68	103	82
## 5	1.84	1	1	7	8	9	8	10	66	108	86
## 6	3.90	1	2	10	10	10	9	9	74	122	98
## 7	2.84	2	1	10	9	10	10	10	63	112	90
## 8	3.57	1	2	10	9	10	10	10	71	120	96
## 9	3.95	2	2	10	10	10	10	9	74	123	98
## 10	3.49	2	1	10	10	9	10	10	75	124	99
## 11	2.32	1	1	7	8	6	7	10	59	97	78
## 12	2.35	1	2	8	10	10	10	9	71	118	94
## 13	2.45	1	1	8	8	10	10	6	69	111	89
## 14	2.90	1	1	3	8	4	6	8	55	84	67
## 15	2.33	1	2	5	5	7	6	4	52	79	63
## 16	2.54	1	2	5	8	6	4	10	61	94	75
## 17	1.66	2	2	5	7	4	7	6	63	92	74
## 18	2.74	1	1	8	9	6	7	10	48	88	70
## 19	2.51	1	1	5	9	5	6	10	63	98	78
## 20	2.54	1	1	10	9	10	10	7	60	106	85
## 21	3.70	1	2	3	6	2	6	6	55	78	62
## 22	3.90	1	2	10	8	10	10	8	68	114	91
## 23	2.21	1	2	9	10	9	9	9	52	98	78
## 24	2.46	1	2	6	9	8	9	9	68	109	87
## 25	1.14	1	2	2	5	4	5	6	43	65	52
## 26	2.09	2	2	6	5	4	7	6	62	90	72
## 27	2.34	2	1	10	8	10	10	7	63	108	86
## 28	2.03	1	2	10	10	10	10	9	71	120	96
## 29	2.80	1	2	4	6	5	4	5	57	81	65
## 30	2.25	2	2	10	9	10	10	8	61	118	86
## 31	2.28	1	2	6	7	9	6	8	61	97	78
## 32	3.02	2	1	10	10	10	9	9	55	103	82
## 33	2.38	1	2	8	9	10	10	9	50	96	77
## 34	1.91	1	2	9	8	10	10	6	66	109	87
## 35	3.53	1	2	6	7	7	9	9	54	92	74
## 36	2.61	1	2	10	10	10	10	10	53	103	82
## 37	2.80	1	1	7	6	9	8	8	60	98	78
## 38	2.66	1	2	8	6	7	8	7	60	96	77
## 39	2.77	1	1	6	8	9	5	8	63	99	79

## 40	3.84	2	2	10	10	10	9	10	74	123	98
## 41	2.38	1	1	4	7	6	4	7	56	84	67
## 42	2.34	1	1	7	6	8	7	9	59	96	77
## 43	3.17	1	2	8	8	8	10	9	70	113	90
## 44	3.42	2	2	10	10	10	9	10	75	124	99
## 45	2.72	1	2	8	9	9	8	10	60	104	83
## 46	2.47	1	1	0	5	0	2	5	40	52	42
## 47	1.33	1	2	8	5	6	4	7	58	88	70
## 48	3.90	1	1	7	8	8	6	6	63	98	78
## 49	3.90	1	1	10	9	10	10	8	74	121	97
## 50	2.96	1	1	7	7	6	9	8	61	98	78
## 51	3.90	2	2	10	10	10	10	8	75	123	98
## 52	3.35	1	1	7	8	8	9	6	69	107	86
## 53	3.84	1	1	4	5	6	6	6	48	75	60
## 54	3.45	2	1	10	8	7	9	7	68	109	87
## 55	3.57	1	2	0	3	2	2	2	42	51	41
## 56	3.05	1	2	9	8	10	8	8	65	108	86
## 57	1.24	1	1	7	6	7	10	5	53	88	70
## 58	1.77	1	2	6	7	6	8	6	59	92	74
## 59	3.64	1	2	10	10	10	10	10	72	122	98
## 60	2.56	1	2	5	7	6	5	6	58	87	70
## 61	1.24	1	2	3	8	5	2	7	59	84	67
## 62	1.50	1	2	5	7	8	5	8	57	90	72
## 63	3.90	1	2	8	9	9	10	10	67	113	90
## 64	3.58	1	2	10	9	10	10	7	62	108	86
## 65	1.61	1	1	6	9	9	7	10	64	105	84
## 66	1.77	1	2	6	7	7	7	5	60	92	74
## 67	2.57	1	1	6	8	9	5	7	62	97	78
## 68	2.96	1	2	8	9	9	9	8	62	105	84
## 69	3.13	1	2	9	6	8	7	10	66	106	85
## 70	3.90	1	2	8	7	9	10	8	57	99	79
## 71	2.27	2	2	10	9	10	10	10	62	111	89
## 72	2.43	1	2	8	8	7	8	10	65	106	85
## 73	2.88	1	1	7	10	8	9	10	60	104	83
## 74	2.40	1	1	3	6	3	2	6	50	70	56
## 75	3.01	1	2	3	5	3	2	4	49	66	53
## 76	3.90	1	2	10	10	10	9	10	73	122	98
## 77	4.00	1	2	9	9	10	10	9	70	117	94
## 78	2.22	2	2	10	9	10	10	8	62	109	87
## 79	2.56	2	1	8	5	6	4	7	62	92	74
## 80	1.95	1	2	9	10	10	8	8	53	98	78
## 81	3.90	1	2	8	9	10	10	8	66	111	89
## 82	3.54	1	2	9	9	10	8	9	68	113	90
## 83	3.66	2	1	10	8	10	10	10	68	116	93
## 84	2.56	1	1	7	6	10	8	7	57	95	76
## 85	2.92	1	2	8	9	8	8	7	68	108	86
## 86	3.90	1	2	8	10	10	8	9	73	118	94
## 87	2.31	1	1	10	8	9	10	7	49	93	74
## 88	2.77	2	2	10	9	10	10	9	61	109	87

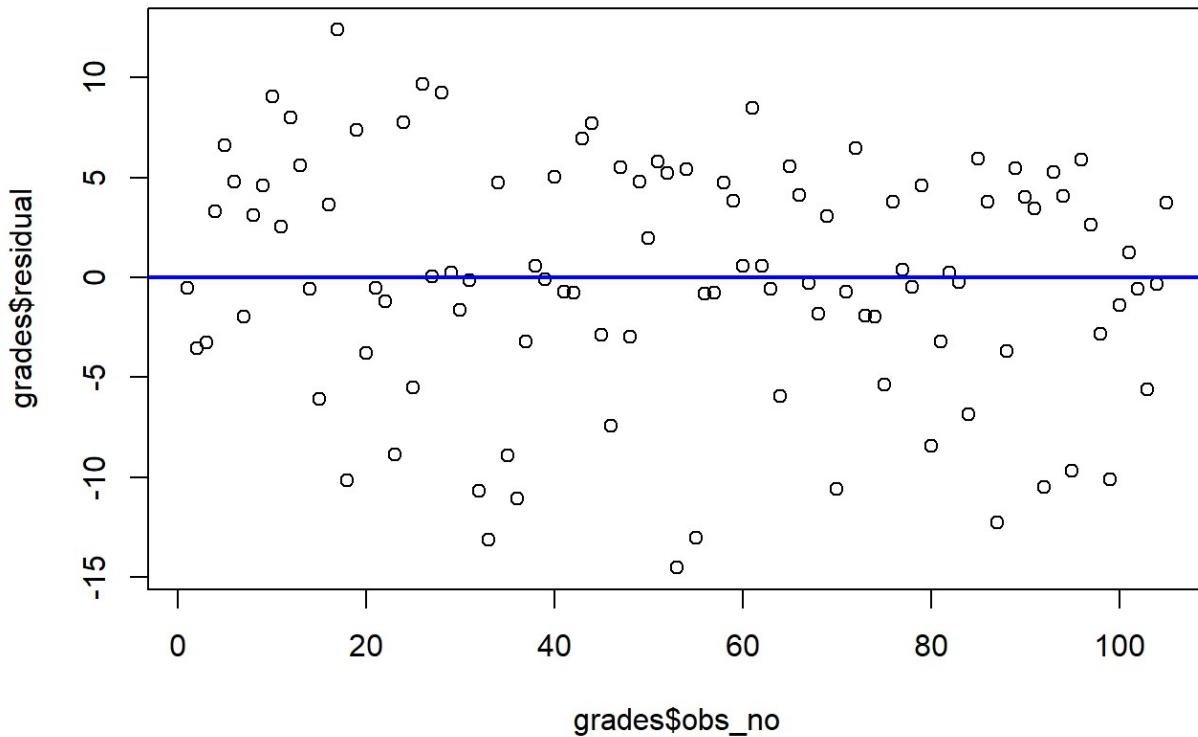
## 89	2.84	1	1	6	8	6	6	10	64	100	80
## 90	1.85	1	2	4	8	5	7	9	57	90	72
## 91	3.49	1	2	10	9	10	10	8	71	118	94
## 92	2.22	1	2	10	10	10	9	8	52	99	79
## 93	3.28	2	2	10	9	10	10	9	72	120	96
## 94	2.73	1	2	7	9	9	7	8	67	107	86
## 95	2.02	2	2	10	8	10	10	7	52	97	78
## 96	3.53	1	2	7	10	9	8	9	72	115	92
## 97	3.90	1	2	6	7	7	5	9	67	101	81
## 98	2.30	1	2	10	9	10	10	7	60	106	85
## 99	2.37	1	2	10	10	10	10	10	53	93	74
## 100	3.19	1	2	9	7	10	9	7	65	107	86
## 101	3.90	1	2	4	8	6	6	9	64	97	78
## 102	1.80	1	2	3	6	3	4	5	49	70	56
## 103	2.11	1	1	4	5	6	6	6	50	77	62
## 104	2.43	2	2	10	9	10	10	7	63	109	87
## 105	3.06	1	2	7	8	9	8	7	68	107	86
## grade passfail residual obs_no											
## 1	D	P	-0.51437366	1							
## 2	C	P	-3.54694323	2							
## 3	C	P	-3.23375140	3							
## 4	B	P	3.30622554	4							
## 5	B	P	6.63289284	5							
## 6	A	P	4.79925345	6							
## 7	A	P	-1.96854482	7							
## 8	A	P	3.11682569	8							
## 9	A	P	4.59962129	9							
## 10	A	P	9.04503165	10							
## 11	C	P	2.54280767	11							
## 12	A	P	7.98785032	12							
## 13	B	P	5.58858601	13							
## 14	D	P	-0.55533632	14							
## 15	D	P	-6.10591327	15							
## 16	C	P	3.66442618	16							
## 17	C	P	12.39554117	17							
## 18	C	P	-10.13410245	18							
## 19	C	P	7.39299999	19							
## 20	B	P	-3.77075188	20							
## 21	D	P	-0.53186180	21							
## 22	A	P	-1.20074655	22							
## 23	C	P	-8.84438513	23							
## 24	B	P	7.76624860	24							
## 25	F	F	-5.52828439	25							
## 26	C	P	9.67870462	26							
## 27	B	P	0.02777675	27							
## 28	A	P	9.26549612	28							
## 29	D	P	0.23513348	29							
## 30	B	P	-1.61288537	30							
## 31	C	P	-0.12387015	31							

## 32	B	P	-10.68722058	32
## 33	C	P	-13.13192897	33
## 34	B	P	4.74461330	34
## 35	C	P	-8.89708504	35
## 36	B	P	-11.05023690	36
## 37	C	P	-3.20004458	37
## 38	C	P	0.57651449	38
## 39	C	P	-0.08026529	39
## 40	A	P	5.03881204	40
## 41	D	P	-0.69675092	41
## 42	C	P	-0.75463422	42
## 43	A	P	6.93147197	43
## 44	A	P	7.71572216	44
## 45	B	P	-2.88063313	45
## 46	F	F	-7.40332171	46
## 47	C	P	5.49552438	47
## 48	C	P	-2.98315752	48
## 49	A	P	4.79925345	49
## 50	C	P	1.98751606	50
## 51	A	P	5.79925345	51
## 52	B	P	5.21279621	52
## 53	D	P	-14.52600990	53
## 54	B	P	5.42232641	54
## 55	F	F	-13.01281820	55
## 56	B	P	-0.80699988	56
## 57	C	P	-0.75393225	57
## 58	C	P	4.73876140	58
## 59	A	P	3.83734067	59
## 60	C	P	0.58457332	60
## 61	D	P	8.46365678	61
## 62	C	P	0.59918602	62
## 63	A	P	-0.59195203	63
## 64	B	P	-5.92310074	64
## 65	B	P	5.55120076	65
## 66	C	P	4.12996688	66
## 67	C	P	-0.28173666	67
## 68	B	P	-1.83886748	68
## 69	B	P	3.09117770	69
## 70	C	P	-10.59195203	70
## 71	B	P	-0.69273823	71
## 72	B	P	6.49482241	72
## 73	B	P	-1.91066152	73
## 74	F	F	-1.95022024	74
## 75	F	F	-5.38573255	75
## 76	A	P	3.79925345	76
## 77	A	P	0.39998914	77
## 78	B	P	-0.49310607	78
## 79	C	P	4.58457332	79
## 80	C	P	-8.41509242	80

```
## 81     B      P -3.20074655    81
## 82     A      O  0.23660498    82
## 83     A      P -0.24251219    83
## 84     C      P -6.85060474    84
## 85     B      P  5.92963276    85
## 86     A      P  3.79925345    86
## 87     C      P -12.24364944   87
## 88     B      P -3.68905980    88
## 89     B      P  5.46663324    89
## 90     C      P  4.02814446    90
## 91     A      P  3.43623714   91
## 92     C      P -10.49310607   92
## 93     A      P  5.27469220   93
## 94     B      P  4.07944044   94
## 95     C      P -9.69457744   95
## 96     A      P  5.88532593   96
## 97     B      P  2.62563700   97
## 98     B      P -2.81251752   98
## 99     C      P -10.09200254  99
## 100    B      P -1.36596992  100
## 101    C      P  1.23443151  101
## 102    F      F -0.55463435  102
## 103    D      P -5.61873727  103
## 104    B      P -0.33156113  104
## 105    B      P  3.76186820  105
```

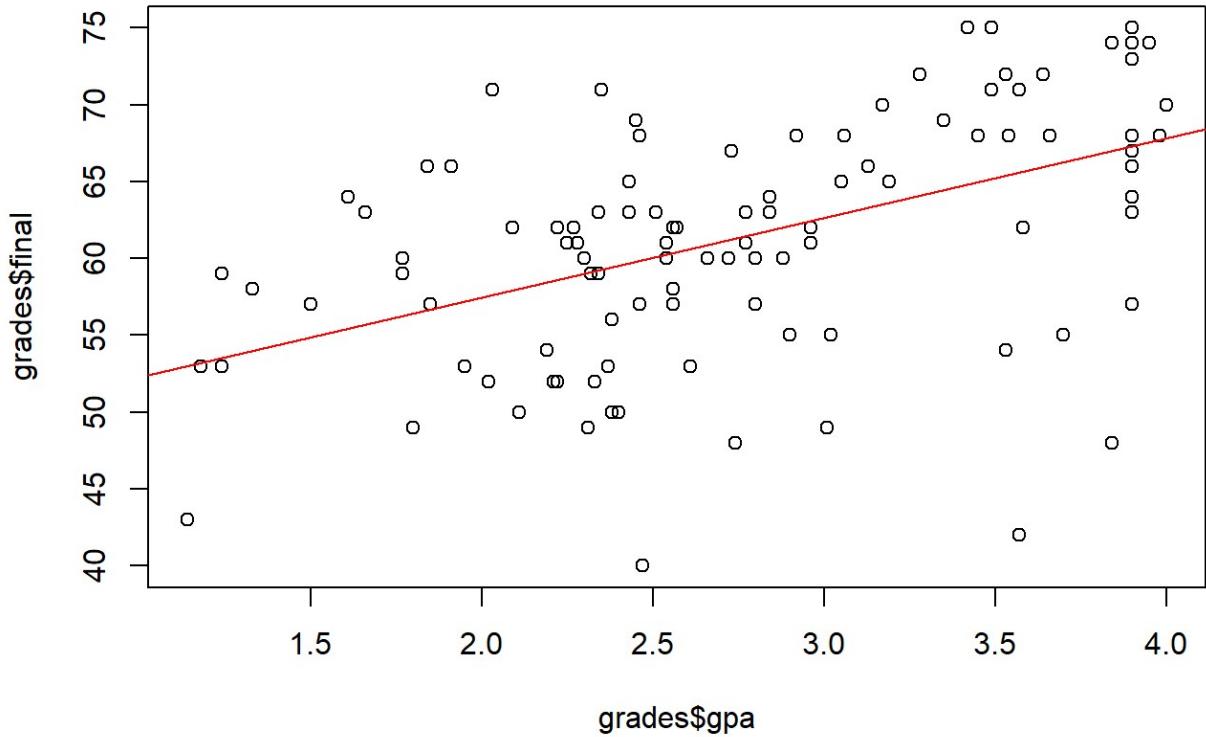
```
#lets plot independent obseration numbers vs error
plot(grades$obs_no, grades$residual, main="Independence of Residuals")
abline(h = 0, col = "blue", lwd=2)
```

Independence of Residuals

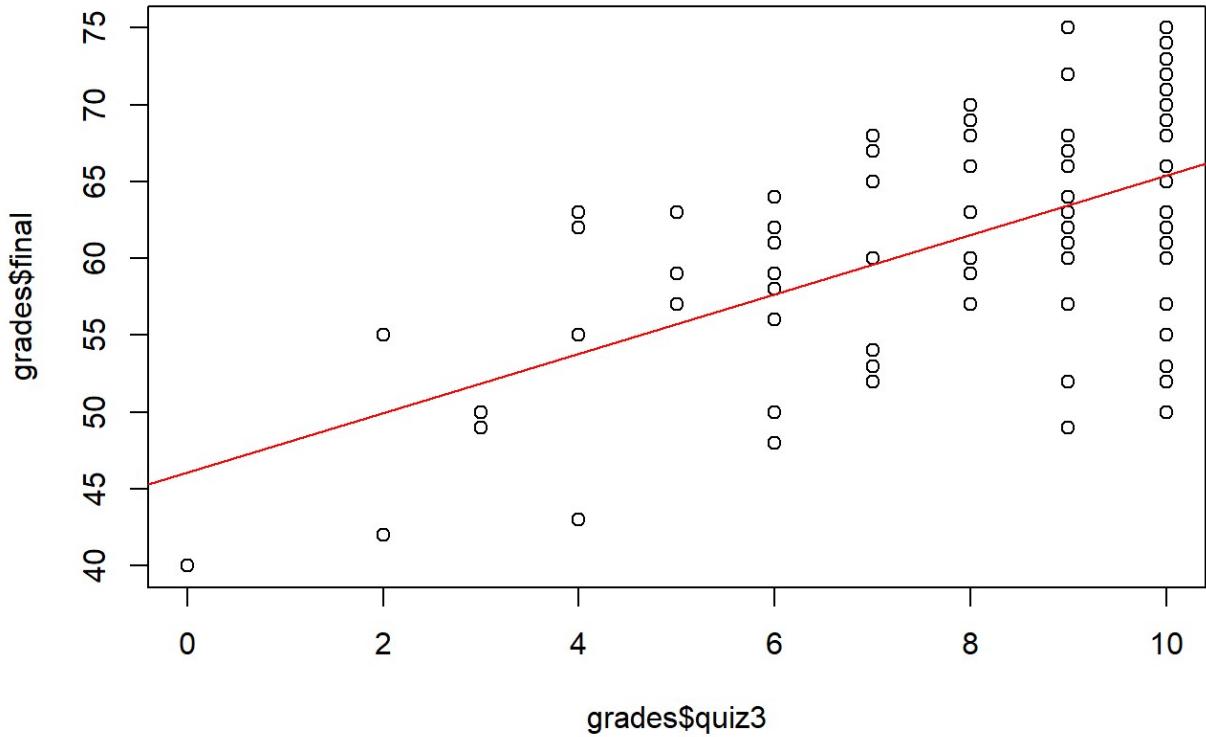


#11. Test the assumption of linear relationship and interpret in maximum 3 lines for each predictor [hint: draw scatter plot between response variable, final (y-axis) and predictor/s (x-axis). If more than one predictor is used in model then more scatter plots would be required]

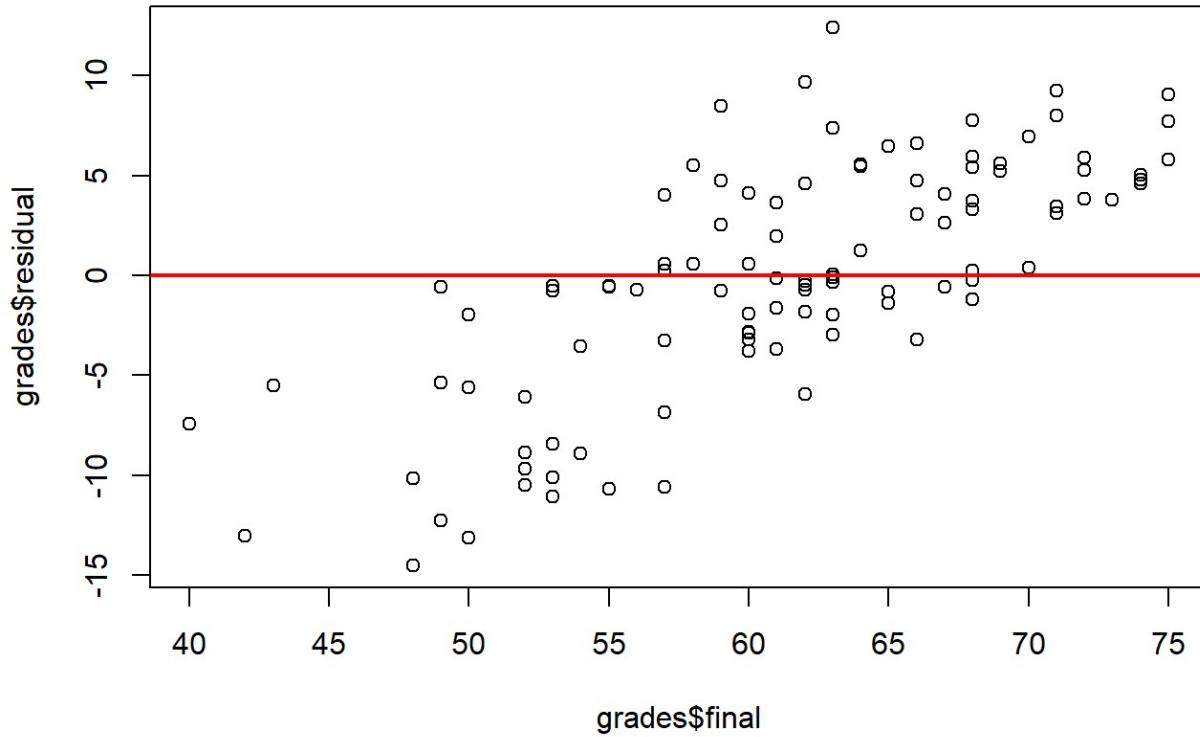
```
plot(grades$gpa, grades$final)
abline(lm(grades$final~grades$gpa), col="red")
```



```
plot(grades$quiz3, grades$final)
abline(lm(grades$final~grades$quiz3), col="red")
```



```
plot(grades$final, grades$residual)
abline(h = 0, col = "red", lwd=2)
```



#12. Test the assumption of Constant Error Variance and interpret in maximum 3 lines [hint: draw scatter plot between residuals/errors (y-axis) and predicted values (x-axis)]

```
pred<-predict(model4)
grades$pred<-NULL
grades$pred<-pred
grades
```

##	Sr_No	id	lastname	firstname	gender	ethnicity	year	lowup	section
## 1	1	106484	VILLARRUZ	ALFRED	2		2	2	1
## 2	2	108642	VALAZQUEZ	SCOTT	2		4	3	2
## 3	3	127285	GALVEZ	JACKIE	1		4	4	2
## 4	4	132931	OSBORNE	ANN	1		3	2	1
## 5	5	140219	GUADIZ	VALERIE	1		2	4	2
## 6	6	142630	RANGIFO	TANIECE	1		4	3	2
## 7	7	153964	TOMOSAWA	DANIEL	2		2	3	2
## 8	8	154441	LIAN	JENNY	1		5	2	1
## 9	9	157147	BAKKEN	KREG	2		4	3	2
## 10	10	164605	LANGFORD	DAWN	1		3	3	2
## 11	11	164842	VALENZUELA	NANCY	1		1	4	2
## 12	12	167664	SWARM	MARK	2		4	3	2
## 13	13	175325	KHOURY	DENNIS	2		4	3	2
## 14	14	192627	MISCHKE	ELAINE	1		4	1	1
## 15	15	211239	AUSTIN	DERRICK	2		4	3	2
## 16	16	219593	POTTER	MICKEY	1		5	3	2
## 17	17	237983	LEE	JONATHAN	2		2	4	2
## 18	18	245473	DAYES	ROBERT	2		4	3	2
## 19	19	249586	STOLL	GLENDON	2		4	3	2
## 20	20	260983	CUSTER	JAMES	2		4	4	2
## 21	21	273611	WU	VIDYUTH	1		2	2	1
## 22	22	280440	CHANG	RENE	1		2	3	2
## 23	23	287617	CUMMINGS	DAVENA	1		5	3	2
## 24	24	289652	BRADLEY	SHANNON	1		4	3	2
## 25	25	302400	JONES	ROBERT	2		3	4	2
## 26	26	307894	TORRENCE	GWEN	1		3	2	1
## 27	27	337908	UYEYAMA	VICTORINE	1		1	3	2
## 28	28	354601	CARPIO	MARY	1		2	2	1
## 29	29	378446	SAUNDERS	TAMARA	1		1	2	1
## 30	30	380157	LUTZ	WILLIAM	2		4	3	2
## 31	31	390203	SHIMA	MIHAELA	1		2	3	2
## 32	32	392464	DOMINGO	MONIKA	1		4	3	2
## 33	33	414775	RATANA	JASON	2		2	3	2
## 34	34	417003	EVANGELIST	NIKKI	1		2	3	2
## 35	35	419891	DE CANIO	PAULA	1		4	3	2
## 36	36	420327	BADGER	SUZANNA	1		4	3	2
## 37	37	434571	SURI	MATHEW	2		2	3	2
## 38	38	436413	PANG	SUZANNE	1		2	3	2
## 39	39	447659	GALANVILLE	DANA	1		5	4	2
## 40	40	463276	HANSEN	TIM	2		4	3	2
## 41	41	466407	PICKERING	HEIDI	1		3	3	2
## 42	42	467806	DEVERS	GAIL	1		3	3	2
## 43	43	473303	PARK	SANDRA	1		3	4	2
## 44	44	479547	LANGFORD	BLAIR	2		3	3	2
## 45	45	490016	STEPHEN	LIZA	1		5	3	2
## 46	46	498900	HUANG	<td>2</td> <td></td> <td>5</td> <td>3</td> <td>2</td>	2		5	3	2
## 47	47	506467	SCARBROUGH	CYNTHE	1		4	3	2

## 48	48	515586	FIALLOS	LAUREL	1	4	2	1	2
## 49	49	519444	RATHBUN	DAWNE	1	4	4	2	2
## 50	50	546022	HAMIDI	KIMBERLY	1	5	3	2	1
## 51	51	553919	KWON	SHELLY	1	2	3	2	1
## 52	52	554809	JONES	LISA	1	3	3	2	3
## 53	53	574170	HURRIA	WAYNE	2	1	2	1	2
## 54	54	576008	BULMERKA	HUSIBA	1	4	4	2	3
## 55	55	576141	MISHALANY	LUCY	1	4	3	2	1
## 56	56	594463	CRUZADO	MARITESS	1	4	4	2	2
## 57	57	595177	WILLIAMS	OLIMPIA	1	3	3	2	3
## 58	58	615115	VASENIUS	RUSS	2	3	3	2	3
## 59	59	616095	SPRINGER	ANNELIES	1	4	3	2	1
## 60	60	623857	CORTEZ	VIKKI	1	3	4	2	3
## 61	61	664653	KHAN	JOHN	2	4	3	2	3
## 62	62	681855	GRISWOLD	TAMMY	1	4	3	2	2
## 63	63	700978	WEBSTER	DEANNA	1	3	2	1	3
## 64	64	703740	SUNYA	DALE	2	5	3	2	3
## 65	65	721311	SONG	LOIS	2	2	3	2	3
## 66	66	725987	BATILLER	FRED	2	2	2	1	2
## 67	67	737728	BELTRAN	JIM	2	3	3	2	1
## 68	68	755724	LANGFORD	TREVOR	2	4	3	2	2
## 69	69	756097	KURSEE	JACKIE	1	3	3	2	2
## 70	70	762308	GOUW	BONNIE	1	4	2	1	3
## 71	71	762813	DAEL	IVAN	2	3	2	1	1
## 72	72	765360	ROBINSON	ERIC	2	3	3	2	2
## 73	73	768995	DUMITRESCU	STACY	2	4	4	2	2
## 74	74	777683	ANDERSON	ERIC	2	5	4	2	3
## 75	75	779481	AHGHEL	BRENDA	1	5	3	2	1
## 76	76	780028	ROBINSON	CLAYTON	2	4	3	2	1
## 77	77	781676	WATKINS	YVONNE	1	3	4	2	1
## 78	78	798931	ZUILL	RENAE	1	4	3	2	1
## 79	79	807963	LEWIS	CARL	2	3	2	1	1
## 80	80	818528	CARRINGTON	JYLL	1	4	3	2	1
## 81	81	822485	VALENZUELA	KATHRYN	1	4	1	1	1
## 82	82	843472	PRADO	DON	2	5	3	2	3
## 83	83	870810	REYNO	NICHOLAS	2	4	3	2	3
## 84	84	896972	HUANG	MIRNA	1	2	3	2	1
## 85	85	897606	GENOBAGA	JACQUELINE	1	2	3	2	3
## 86	86	898766	RAO	DAWN	1	2	3	2	1
## 87	87	899529	HAWKINS	CARHERINE	1	3	4	2	2
## 88	88	900485	COCHRAN	STACY	2	4	3	2	2
## 89	89	905109	JENKINS	ERIC	2	3	2	1	3
## 90	90	908754	MARQUEZ	CHYRELLE	1	4	1	1	2
## 91	91	911355	LESKO	LETICIA	1	3	2	1	3
## 92	92	915457	SHEARER	LUCIO	2	3	3	2	1
## 93	93	920656	LIAO	MICHELLE	1	2	2	1	2
## 94	94	921297	KINZER	RICHARD	2	4	3	2	2
## 95	95	938666	SUAREZ-TAN	KHANH	1	2	3	2	3
## 96	96	938881	YEO	DENISE	1	1	3	2	3

## 97	97	944702	LEDESMA	MARTINE	1	4	3	2	2		
## 98	98	958384	RONCO	SHERRY	1	4	2	1	1		
## 99	99	972678	KAHRS	JANN	1	4	4	2	2		
## 100	100	973427	ROSS	MARIA	1	4	4	2	1		
## 101	101	978889	ZIMCHEK	ARMANDO	2	4	4	2	1		
## 102	102	979028	NEUHARTH	JIM	2	4	3	2	3		
## 103	103	983522	SLOAT	AARON	2	3	3	2	3		
## 104	104	985700	CHA	LILY	1	4	2	1	1		
## 105	105	988808	MCCONAHA	CORA	1	4	3	2	3		
## gpa extrc review quiz1 quiz2 quiz3 quiz4 quiz5 final total percent											
## 1	1.18	1	2	6	5	7	6	3	53	80	64
## 2	2.19	2	1	10	10	7	6	9	54	96	77
## 3	2.46	2	2	10	7	8	9	7	57	98	78
## 4	3.98	1	1	7	8	7	7	6	68	103	82
## 5	1.84	1	1	7	8	9	8	10	66	108	86
## 6	3.90	1	2	10	10	10	9	9	74	122	98
## 7	2.84	2	1	10	9	10	10	10	63	112	90
## 8	3.57	1	2	10	9	10	10	10	71	120	96
## 9	3.95	2	2	10	10	10	10	9	74	123	98
## 10	3.49	2	1	10	10	9	10	10	75	124	99
## 11	2.32	1	1	7	8	6	7	10	59	97	78
## 12	2.35	1	2	8	10	10	10	9	71	118	94
## 13	2.45	1	1	8	8	10	10	6	69	111	89
## 14	2.90	1	1	3	8	4	6	8	55	84	67
## 15	2.33	1	2	5	5	7	6	4	52	79	63
## 16	2.54	1	2	5	8	6	4	10	61	94	75
## 17	1.66	2	2	5	7	4	7	6	63	92	74
## 18	2.74	1	1	8	9	6	7	10	48	88	70
## 19	2.51	1	1	5	9	5	6	10	63	98	78
## 20	2.54	1	1	10	9	10	10	7	60	106	85
## 21	3.70	1	2	3	6	2	6	6	55	78	62
## 22	3.90	1	2	10	8	10	10	8	68	114	91
## 23	2.21	1	2	9	10	9	9	9	52	98	78
## 24	2.46	1	2	6	9	8	9	9	68	109	87
## 25	1.14	1	2	2	5	4	5	6	43	65	52
## 26	2.09	2	2	6	5	4	7	6	62	90	72
## 27	2.34	2	1	10	8	10	10	7	63	108	86
## 28	2.03	1	2	10	10	10	10	9	71	120	96
## 29	2.80	1	2	4	6	5	4	5	57	81	65
## 30	2.25	2	2	10	9	10	10	8	61	118	86
## 31	2.28	1	2	6	7	9	6	8	61	97	78
## 32	3.02	2	1	10	10	10	9	9	55	103	82
## 33	2.38	1	2	8	9	10	10	9	50	96	77
## 34	1.91	1	2	9	8	10	10	6	66	109	87
## 35	3.53	1	2	6	7	7	9	9	54	92	74
## 36	2.61	1	2	10	10	10	10	10	53	103	82
## 37	2.80	1	1	7	6	9	8	8	60	98	78
## 38	2.66	1	2	8	6	7	8	7	60	96	77
## 39	2.77	1	1	6	8	9	5	8	63	99	79

## 40	3.84	2	2	10	10	10	9	10	74	123	98
## 41	2.38	1	1	4	7	6	4	7	56	84	67
## 42	2.34	1	1	7	6	8	7	9	59	96	77
## 43	3.17	1	2	8	8	8	10	9	70	113	90
## 44	3.42	2	2	10	10	10	9	10	75	124	99
## 45	2.72	1	2	8	9	9	8	10	60	104	83
## 46	2.47	1	1	0	5	0	2	5	40	52	42
## 47	1.33	1	2	8	5	6	4	7	58	88	70
## 48	3.90	1	1	7	8	8	6	6	63	98	78
## 49	3.90	1	1	10	9	10	10	8	74	121	97
## 50	2.96	1	1	7	7	6	9	8	61	98	78
## 51	3.90	2	2	10	10	10	10	8	75	123	98
## 52	3.35	1	1	7	8	8	9	6	69	107	86
## 53	3.84	1	1	4	5	6	6	6	48	75	60
## 54	3.45	2	1	10	8	7	9	7	68	109	87
## 55	3.57	1	2	0	3	2	2	2	42	51	41
## 56	3.05	1	2	9	8	10	8	8	65	108	86
## 57	1.24	1	1	7	6	7	10	5	53	88	70
## 58	1.77	1	2	6	7	6	8	6	59	92	74
## 59	3.64	1	2	10	10	10	10	10	72	122	98
## 60	2.56	1	2	5	7	6	5	6	58	87	70
## 61	1.24	1	2	3	8	5	2	7	59	84	67
## 62	1.50	1	2	5	7	8	5	8	57	90	72
## 63	3.90	1	2	8	9	9	10	10	67	113	90
## 64	3.58	1	2	10	9	10	10	7	62	108	86
## 65	1.61	1	1	6	9	9	7	10	64	105	84
## 66	1.77	1	2	6	7	7	7	5	60	92	74
## 67	2.57	1	1	6	8	9	5	7	62	97	78
## 68	2.96	1	2	8	9	9	9	8	62	105	84
## 69	3.13	1	2	9	6	8	7	10	66	106	85
## 70	3.90	1	2	8	7	9	10	8	57	99	79
## 71	2.27	2	2	10	9	10	10	10	62	111	89
## 72	2.43	1	2	8	8	7	8	10	65	106	85
## 73	2.88	1	1	7	10	8	9	10	60	104	83
## 74	2.40	1	1	3	6	3	2	6	50	70	56
## 75	3.01	1	2	3	5	3	2	4	49	66	53
## 76	3.90	1	2	10	10	10	9	10	73	122	98
## 77	4.00	1	2	9	9	10	10	9	70	117	94
## 78	2.22	2	2	10	9	10	10	8	62	109	87
## 79	2.56	2	1	8	5	6	4	7	62	92	74
## 80	1.95	1	2	9	10	10	8	8	53	98	78
## 81	3.90	1	2	8	9	10	10	8	66	111	89
## 82	3.54	1	2	9	9	10	8	9	68	113	90
## 83	3.66	2	1	10	8	10	10	10	68	116	93
## 84	2.56	1	1	7	6	10	8	7	57	95	76
## 85	2.92	1	2	8	9	8	8	7	68	108	86
## 86	3.90	1	2	8	10	10	8	9	73	118	94
## 87	2.31	1	1	10	8	9	10	7	49	93	74
## 88	2.77	2	2	10	9	10	10	9	61	109	87

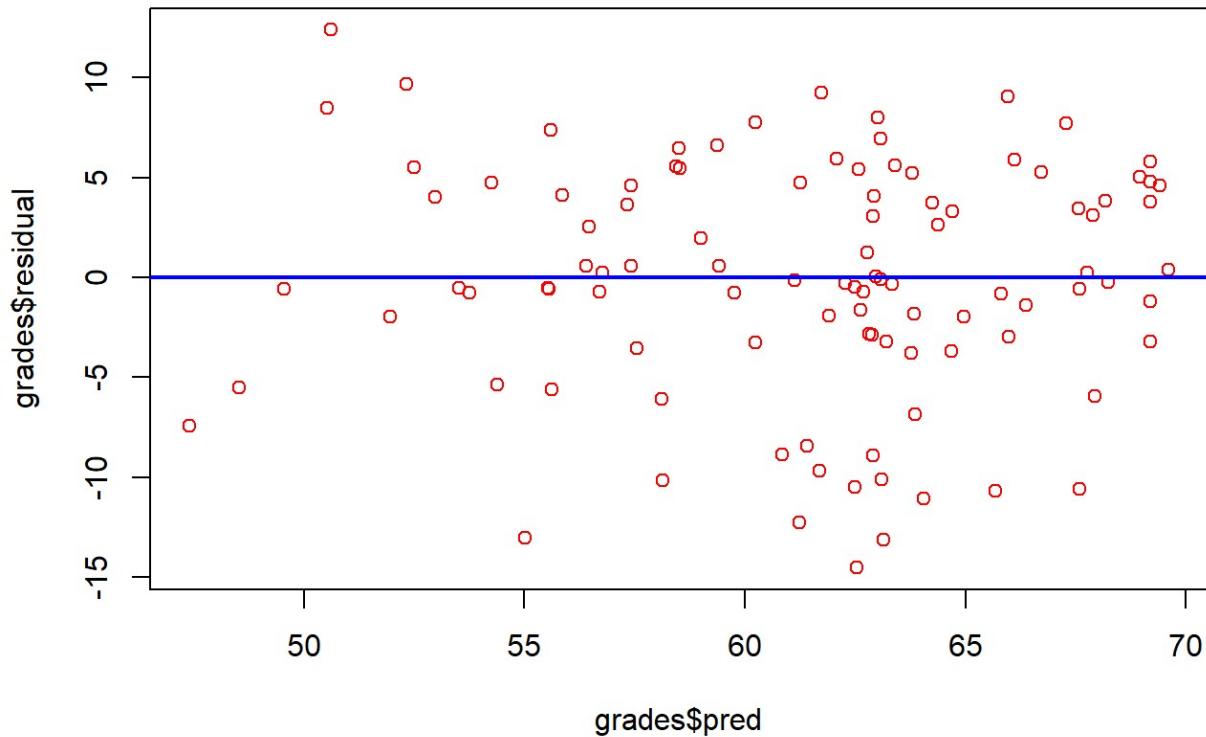
## 89	2.84	1	1	6	8	6	6	10	64	100	80
## 90	1.85	1	2	4	8	5	7	9	57	90	72
## 91	3.49	1	2	10	9	10	10	8	71	118	94
## 92	2.22	1	2	10	10	10	9	8	52	99	79
## 93	3.28	2	2	10	9	10	10	9	72	120	96
## 94	2.73	1	2	7	9	9	7	8	67	107	86
## 95	2.02	2	2	10	8	10	10	7	52	97	78
## 96	3.53	1	2	7	10	9	8	9	72	115	92
## 97	3.90	1	2	6	7	7	5	9	67	101	81
## 98	2.30	1	2	10	9	10	10	7	60	106	85
## 99	2.37	1	2	10	10	10	10	10	53	93	74
## 100	3.19	1	2	9	7	10	9	7	65	107	86
## 101	3.90	1	2	4	8	6	6	9	64	97	78
## 102	1.80	1	2	3	6	3	4	5	49	70	56
## 103	2.11	1	1	4	5	6	6	6	50	77	62
## 104	2.43	2	2	10	9	10	10	7	63	109	87
## 105	3.06	1	2	7	8	9	8	7	68	107	86
##	grade	passfail		residual	obs_no			pred			
## 1	D	P	-0.51437366		1	53.51437					
## 2	C	P	-3.54694323		2	57.54694					
## 3	C	P	-3.23375140		3	60.23375					
## 4	B	P	3.30622554		4	64.69377					
## 5	B	P	6.63289284		5	59.36711					
## 6	A	P	4.79925345		6	69.20075					
## 7	A	P	-1.96854482		7	64.96854					
## 8	A	P	3.11682569		8	67.88317					
## 9	A	P	4.59962129		9	69.40038					
## 10	A	P	9.04503165		10	65.95497					
## 11	C	P	2.54280767		11	56.45719					
## 12	A	P	7.98785032		12	63.01215					
## 13	B	P	5.58858601		13	63.41141					
## 14	D	P	-0.55533632		14	55.55534					
## 15	D	P	-6.10591327		15	58.10591					
## 16	C	P	3.66442618		16	57.33557					
## 17	C	P	12.39554117		17	50.60446					
## 18	C	P	-10.13410245		18	58.13410					
## 19	C	P	7.39299999		19	55.60700					
## 20	B	P	-3.77075188		20	63.77075					
## 21	D	P	-0.53186180		21	55.53186					
## 22	A	P	-1.20074655		22	69.20075					
## 23	C	P	-8.84438513		23	60.84439					
## 24	B	P	7.76624860		24	60.23375					
## 25	F	F	-5.52828439		25	48.52828					
## 26	C	P	9.67870462		26	52.32130					
## 27	B	P	0.02777675		27	62.97222					
## 28	A	P	9.26549612		28	61.73450					
## 29	D	P	0.23513348		29	56.76487					
## 30	B	P	-1.61288537		30	62.61289					
## 31	C	P	-0.12387015		31	61.12387					

## 32	B	P	-10.68722058	32	65.68722
## 33	C	P	-13.13192897	33	63.13193
## 34	B	P	4.74461330	34	61.25539
## 35	C	P	-8.89708504	35	62.89709
## 36	B	P	-11.05023690	36	64.05024
## 37	C	P	-3.20004458	37	63.20004
## 38	C	P	0.57651449	38	59.42349
## 39	C	P	-0.08026529	39	63.08027
## 40	A	P	5.03881204	40	68.96119
## 41	D	P	-0.69675092	41	56.69675
## 42	C	P	-0.75463422	42	59.75463
## 43	A	P	6.93147197	43	63.06853
## 44	A	P	7.71572216	44	67.28428
## 45	B	P	-2.88063313	45	62.88063
## 46	F	F	-7.40332171	46	47.40332
## 47	C	P	5.49552438	47	52.50448
## 48	C	P	-2.98315752	48	65.98316
## 49	A	P	4.79925345	49	69.20075
## 50	C	P	1.98751606	50	59.01248
## 51	A	P	5.79925345	51	69.20075
## 52	B	P	5.21279621	52	63.78720
## 53	D	P	-14.52600990	53	62.52601
## 54	B	P	5.42232641	54	62.57767
## 55	F	F	-13.01281820	55	55.01282
## 56	B	P	-0.80699988	56	65.80700
## 57	C	P	-0.75393225	57	53.75393
## 58	C	P	4.73876140	58	54.26124
## 59	A	P	3.83734067	59	68.16266
## 60	C	P	0.58457332	60	57.41543
## 61	D	P	8.46365678	61	50.53634
## 62	C	P	0.59918602	62	56.40081
## 63	A	P	-0.59195203	63	67.59195
## 64	B	P	-5.92310074	64	67.92310
## 65	B	P	5.55120076	65	58.44880
## 66	C	P	4.12996688	66	55.87003
## 67	C	P	-0.28173666	67	62.28174
## 68	B	P	-1.83886748	68	63.83887
## 69	B	P	3.09117770	69	62.90882
## 70	C	P	-10.59195203	70	67.59195
## 71	B	P	-0.69273823	71	62.69274
## 72	B	P	6.49482241	72	58.50518
## 73	B	P	-1.91066152	73	61.91066
## 74	F	F	-1.95022024	74	51.95022
## 75	F	F	-5.38573255	75	54.38573
## 76	A	P	3.79925345	76	69.20075
## 77	A	P	0.39998914	77	69.60001
## 78	B	P	-0.49310607	78	62.49311
## 79	C	P	4.58457332	79	57.41543
## 80	C	P	-8.41509242	80	61.41509

```
## 81     B      P -3.20074655    81 69.20075
## 82     A      O  0.23660498    82 67.76340
## 83     A      P -0.24251219    83 68.24251
## 84     C      P -6.85060474    84 63.85060
## 85     B      P  5.92963276    85 62.07037
## 86     A      P  3.79925345    86 69.20075
## 87     C      P -12.24364944   87 61.24365
## 88     B      P -3.68905980   88 64.68906
## 89     B      P  5.46663324   89 58.53337
## 90     C      P  4.02814446   90 52.97186
## 91     A      P  3.43623714   91 67.56376
## 92     C      P -10.49310607  92 62.49311
## 93     A      P  5.27469220   93 66.72531
## 94     B      P  4.07944044   94 62.92056
## 95     C      P -9.69457744  95 61.69458
## 96     A      P  5.88532593   96 66.11467
## 97     B      P  2.62563700   97 64.37436
## 98     B      P -2.81251752  98 62.81252
## 99     C      P -10.09200254  99 63.09200
## 100    B      P -1.36596992  100 66.36597
## 101    C      P  1.23443151  101 62.76557
## 102    F      F -0.55463435  102 49.55463
## 103    D      P -5.61873727  103 55.61874
## 104    B      P -0.33156113  104 63.33156
## 105    B      P  3.76186820  105 64.23813
```

```
plot(grades$pred,grades$residual, main = "Constant error Variance", col="red")
abline(h = 0, col = "blue", lwd=2)
```

Constant error Variance



#What is Standard Error of Estimate of your model and how do you interpret the same. Show with some hypothetical values of predictors. Maximum 300 words. [hint: Standard Error of Estimate]
fitted(model4)

```

##      1      2      3      4      5      6      7      8
## 53.51437 57.54694 60.23375 64.69377 59.36711 69.20075 64.96854 67.88317
##      9     10     11     12     13     14     15     16
## 69.40038 65.95497 56.45719 63.01215 63.41141 55.55534 58.10591 57.33557
##     17     18     19     20     21     22     23     24
## 50.60446 58.13410 55.60700 63.77075 55.53186 69.20075 60.84439 60.23375
##     25     26     27     28     29     30     31     32
## 48.52828 52.32130 62.97222 61.73450 56.76487 62.61289 61.12387 65.68722
##     33     34     35     36     37     38     39     40
## 63.13193 61.25539 62.89709 64.05024 63.20004 59.42349 63.08027 68.96119
##     41     42     43     44     45     46     47     48
## 56.69675 59.75463 63.06853 67.28428 62.88063 47.40332 52.50448 65.98316
##     49     50     51     52     53     54     55     56
## 69.20075 59.01248 69.20075 63.78720 62.52601 62.57767 55.01282 65.80700
##     57     58     59     60     61     62     63     64
## 53.75393 54.26124 68.16266 57.41543 50.53634 56.40081 67.59195 67.92310
##     65     66     67     68     69     70     71     72
## 58.44880 55.87003 62.28174 63.83887 62.90882 67.59195 62.69274 58.50518
##     73     74     75     76     77     78     79     80
## 61.91066 51.95022 54.38573 69.20075 69.60001 62.49311 57.41543 61.41509
##     81     82     83     84     85     86     87     88
## 69.20075 67.76340 68.24251 63.85060 62.07037 69.20075 61.24365 64.68906
##     89     90     91     92     93     94     95     96
## 58.53337 52.97186 67.56376 62.49311 66.72531 62.92056 61.69458 66.11467
##     97     98     99    100    101    102    103    104
## 64.37436 62.81252 63.09200 66.36597 62.76557 49.55463 55.61874 63.33156
##     105
## 64.23813

```

R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <http://rmarkdown.rstudio.com> (<http://rmarkdown.rstudio.com>).

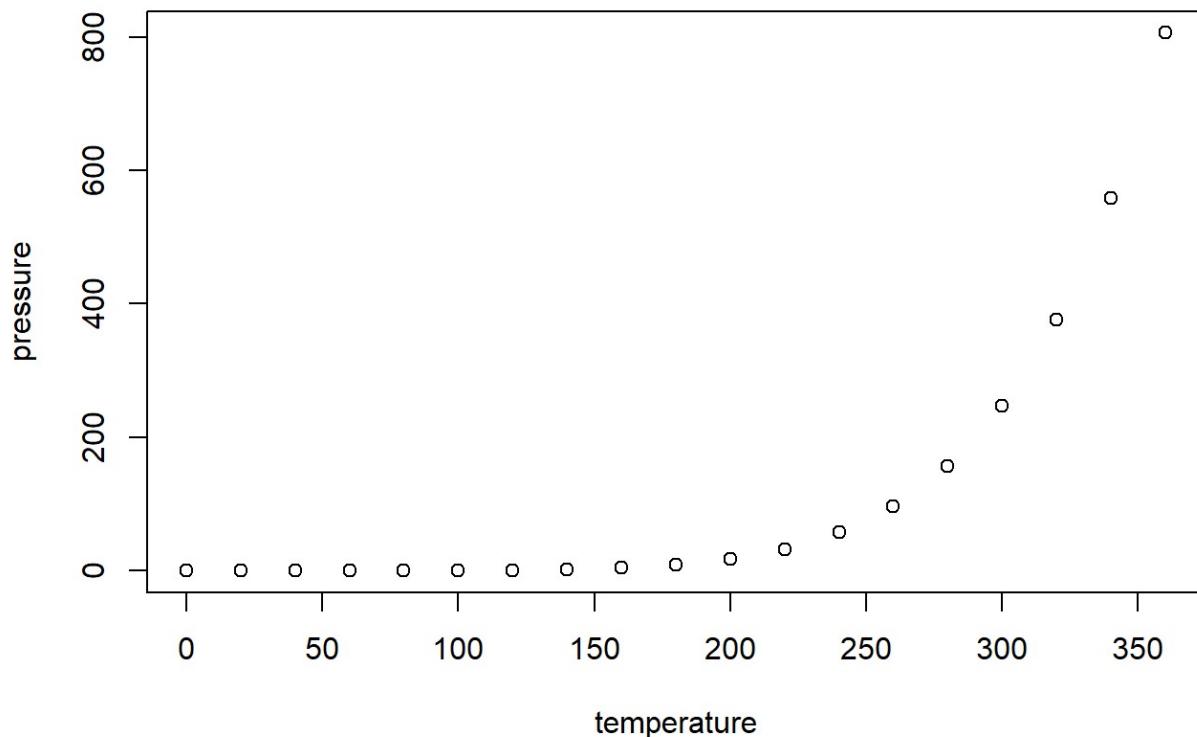
When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
summary(cars)
```

```
##      speed          dist
##  Min.   : 4.0   Min.   : 2.00
##  1st Qu.:12.0  1st Qu.: 26.00
##  Median :15.0  Median : 36.00
##  Mean   :15.4  Mean   : 42.98
##  3rd Qu.:19.0  3rd Qu.: 56.00
##  Max.   :25.0  Max.   :120.00
```

Including Plots

You can also embed plots, for example:



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.