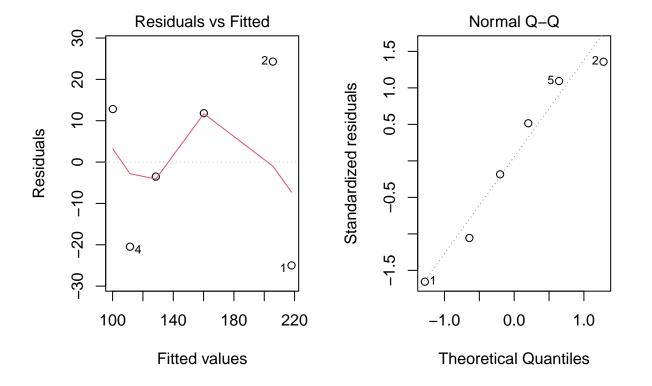
## 101B - HW4

```
#10.6
dat \leftarrow data.frame(y = c(193, 230, 172, 91, 113, 125),
                  x1 = c(1.6, 15.5, 22.0, 43.0, 33.0, 40.0),
                  x2 = c(851,816,1058,1201,1357,1115))
mod \leftarrow lm(y~x1+x2,dat)
summary(mod)
##
## Call:
## lm(formula = y \sim x1 + x2, data = dat)
##
## Residuals:
         1
                 2
                         3
                                          5
## -24.987 24.307 11.820 -20.460 12.830 -3.511
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 350.99427
                            74.75307
                                     4.695
                                               0.0183 *
## x1
                             1.16914 -1.088
                                               0.3562
                -1.27199
## x2
                -0.15390
                             0.08953 -1.719
                                               0.1841
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 25.5 on 3 degrees of freedom
## Multiple R-squared: 0.8618, Adjusted R-squared: 0.7696
## F-statistic: 9.353 on 2 and 3 DF, p-value: 0.05138
#linear reg model = 350.99-1.27x1 - 0.15x2
#Our p-value: 0.051 is greater than a = 0.05, concluding that we reject our null and our regression is
#The given t values are:
#p-value: 0.0188, 0.3562, 0.1841
\#b0 = 4.695, b1 - 1.088, b2 = -0.1539
#We can conclude pvalues 2 and 3 are not significant in this model, but the p1 is significant.
#95% CI X1:
x \leftarrow lm(y\sim x1, dat)
confint(x, 'x1', level = 0.95)
```

```
2.5 %
                     97.5 %
## x1 -5.293793 -0.4181486
z \leftarrow lm(y\sim x2, dat)
confint(z, 'x2', level = 0.95)
##
           2.5 %
                       97.5 %
## x2 -0.3871457 -0.07420014
res.aov <- aov(y~x1+x2,dat)
summary(res.aov)
##
               Df Sum Sq Mean Sq F value Pr(>F)
                            10240
                                  15.751 0.0286 *
## x1
                    10240
## x2
                                     2.955 0.1841
                     1921
                             1921
                 1
## Residuals
                     1950
                              650
## ---
## Signif. codes:
                    0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
par(mfrow=c(1,2))
plot(res.aov,1:2)
```



```
##10.9 #A
11<- c(yield = 1,1,1,1,1,1,1,1,1,</pre>
      concentration = 1,1,2,2,1,1,2,2,
       temperature = 150,180,150,180,150,180,150,180)
m1 <- matrix(l1, nrow = 8, ncol = 3, byrow = FALSE)</pre>
colnames(m1) <- c("y","x1","x2")</pre>
##
       y x1 x2
## [1,] 1 1 150
## [2,] 1 1 180
## [3,] 1 2 150
## [4,] 1 2 180
## [5,] 1 1 150
## [6,] 1 1 180
## [7,] 1 2 150
## [8,] 1 2 180
150,180,150,180,150,180,150,180)
m2 <- matrix(12, nrow = 3, ncol = 8, byrow = TRUE)
m2%*%m1
                     x2
              x1
          У
                   1320
## [1,]
          8
              12
## [2,]
        12
              20
                   1980
## [3,] 1320 1980 219600
#B #The matrix is not a diagonal. We have obtained values other than zero outside of the diagonal.
\#C
m1[,"x2"] \leftarrow (m1[,"x2"] - 165)/15
m1[,"x2"]
## [1] -1 1 -1 1 -1 1 -1 1
m1[,"x1"] \leftarrow (m1[,"x1"] - 1.5)/0.5
m1
       y x1 x2
##
## [1,] 1 -1 -1
## [2,] 1 -1 1
## [3,] 1 1 -1
## [4,] 1 1 1
```

#The normality is fair, but the residual plot is slightly skewed.

```
## [5,] 1 -1 -1
## [6,] 1 -1 1
## [7,] 1 1 -1
## [8,] 1 1 1
inv <- c(1,1,1,1,1,1,1,1,1,1,
           -1, -1, 1, 1, -1, -1, 1, 1,
            -1,1,-1,1,-1,1,-1,1
inv1 <- matrix(inv, nrow = 3, ncol = 8, byrow = TRUE)</pre>
        [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]
                            1
         1 1 1 1
                                 1
## [2,]
                                 -1
         -1
             -1
                   1
                         1
                            -1
                                         1
                                              1
## [3,]
        -1
              1 -1
                            -1
inv1 %*% m1
##
      y x1 x2
## [1,] 8 0 0
## [2,] 0 8 0
## [3,] 0 0 8
#The matrix is a diagonal because all of the values outside of the diagnol are equal to 0.
\#D
11<- c(yield = 1,1,1,1,1,1,1,1,1,</pre>
       concentration = 1,1,2,2,1,1,2,2,
       temperature = 150,180,150,180,150,180,150,180)
m1 <- matrix(l1, nrow = 8, ncol = 3, byrow = FALSE)</pre>
colnames(m1) <- c("y", "x1", "x2")</pre>
m1[,"x2"] \leftarrow (m1[,"x2"] - 150)/30
m1[,"x2"]
## [1] 0 1 0 1 0 1 0 1
m1[,"x1"] \leftarrow (m1[,"x1"] - 1.0)/1.0
##
       y x1 x2
## [1,] 1 0 0
## [2,] 1 0 1
## [3,] 1 1 0
## [4,] 1 1 1
## [5,] 1 0 0
## [6,] 1 0 1
## [7,] 1 1 0
## [8,] 1 1 1
```

```
list2 <- c(1,1,1,1,1,1,1,1,1,1,
        0,0,1,1,0,0,1,1,
        0,1,0,1,0,1,0,1)
m3 <- matrix(list2, nrow = 3, ncol = 8, TRUE)
m3
##
        [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,]
                1
                     1
                          1
                               1
                                     1
                                          1
## [2,]
           0
                0
                               0
                                     0
                                          1
                     1
                          1
                                               1
## [3,]
                     0
                                               1
m3 %*% m1
##
        y x1 x2
## [1,] 8
           4 4
## [2,] 4 4 2
## [3,] 4 2 4
```

#The matrix is not a diagonal. We have obtained values other than zero outside of the diagonal.

#E From theis exercise I have learned that there are many ways to manipulate data and through vectors and matrices. Orthogonal designed matrix are easiest to deal with the orthagonal design.