Linear HW#6

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7.22

There is one possible explaination of this phenanimon, multicolinearity among the predictor variables. Because of multicolinearity, two variable can fit the data fairly well while other coefficients are not statistically significant. But multicolinearity does not inhibit us from obtaining precise estimate value. With more predictor included, the predictions will be more precise.

7.24

```
a)
data7.24<-read.table("CHO6PRO5.txt")
colnames(data7.24)<-c("Y","X1","X2")</pre>
lm7.24a < -lm(Y~X1,data=data7.24)
lm7.24a
##
## Call:
## lm(formula = Y ~ X1, data = data7.24)
##
## Coefficients:
##
   (Intercept)
                           X1
        50.775
                        4.425
##
Y = 50.775 + 4.425 X_1
b)
lm6.5 < -lm(Y \sim X2 + X1, data = data7.24)
lm6.5
##
## lm(formula = Y \sim X2 + X1, data = data7.24)
##
## Coefficients:
                           Х2
                                          Х1
## (Intercept)
                                       4.425
         37.650
                        4.375
cor(data7.24$X1,data7.24$X2)
## [1] 0
```

T1 0

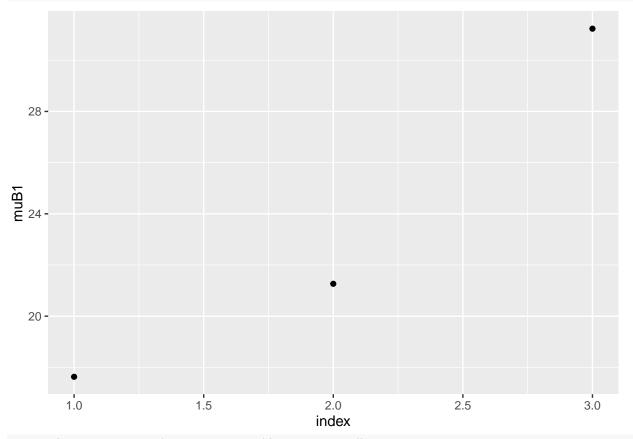
The coefficient β_1 from part a is identical to that of 6.5.

c)

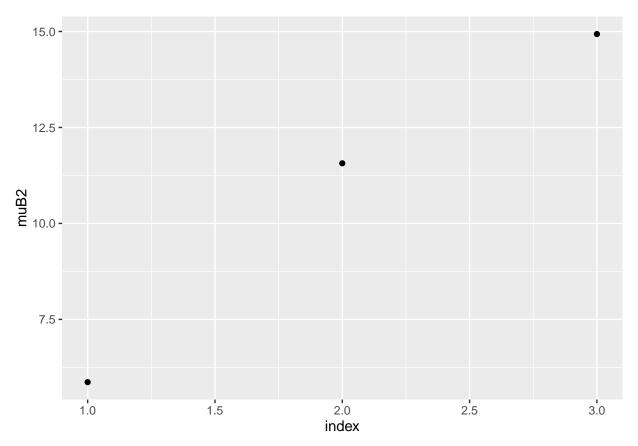
```
anova(lm7.24a)
## Analysis of Variance Table
##
## Response: Y
##
             Df Sum Sq Mean Sq F value
                                             Pr(>F)
## X1
              1 1566.45 1566.45 54.751 3.356e-06 ***
## Residuals 14 400.55
                           28.61
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
anova(lm6.5)
## Analysis of Variance Table
##
## Response: Y
             Df
                 Sum Sq Mean Sq F value
                                             Pr(>F)
                 306.25 306.25 42.219 2.011e-05 ***
## X2
## X1
              1 1566.45 1566.45 215.947 1.778e-09 ***
## Residuals 13
                  94.30
                            7.25
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Yes, SSR(X_1) equals to SSR(X_1|X_2).
d)
The correlation between X1 and X_2 is 0, so the infomation in these two variable does not overlap, the two
models from 6.5 and 7.24 produce the same \beta_1, SSR(X_1) equals to SSR(X_1|X_2).
Problem 3
a)
strength < -c(17.8, 18.2, 16.9, 21.4, 20.1, 22.3, 30.6, 32.1, 31.0, 6.5, 5.2, 5.9, 10.4, 11.7, 12.6, 14.6, 14.3, 15.9)
A \leftarrow c(1,1,1,2,2,2,3,3,3,1,1,1,2,2,2,3,3,3)
B < -c(rep(1,9), rep(2,9))
group<-c(1,1,1,2,2,2,3,3,3,4,4,4,5,5,5,6,6,6)
data3<-data.frame(strength=strength,group=group,A=A,B=B)
lm3.a<-lm(strength~as.factor(group),data=data3)</pre>
lm3.a
##
## Call:
## lm(formula = strength ~ as.factor(group), data = data3)
## Coefficients:
##
         (Intercept)
                      as.factor(group)2 as.factor(group)3
##
              17.633
                                    3.633
                                                       13.600
## as.factor(group)4
                      as.factor(group)5
                                          as.factor(group)6
##
             -11.767
                                   -6.067
                                                       -2.700
library(ggplot2)
muB1<-c(17.633,17.633+3.633,17.633+13.600)
```

muB2<-c(17.633-11.767,17.633-6.067,17.633-2.7)

```
index<-c(1,2,3)
data3.plot<-data.frame(muB1=muB1,muB2=muB2,index=index)
ggplot(data3.plot,aes(x=index,y=muB1))+geom_point()</pre>
```



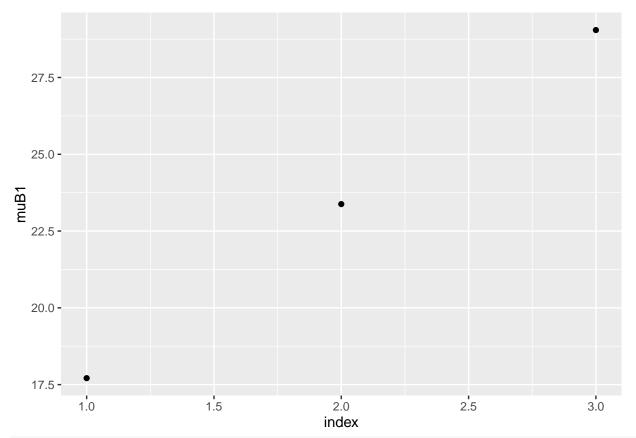
ggplot(data3.plot,aes(x=index,y=muB2))+geom_point()



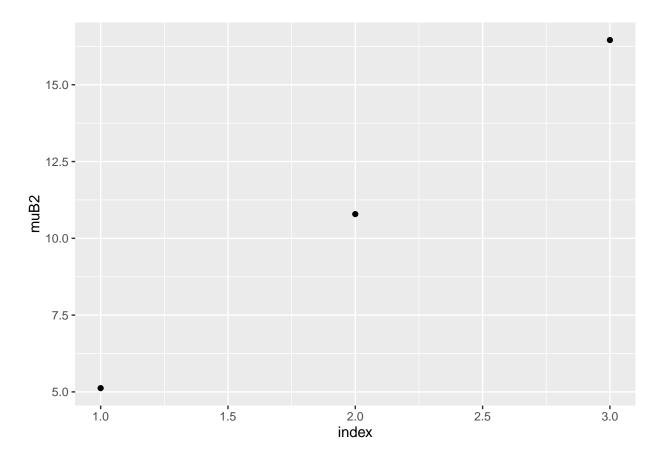
 $\beta_0 = 17.633, \, \beta_1 = 3.633, \, \beta_2 = 13.600, \, \beta_3 = \text{-}11.767, \, \beta_4 = \text{-}6.067, \, \beta_5 = \text{-}2.700.$

```
b)
```

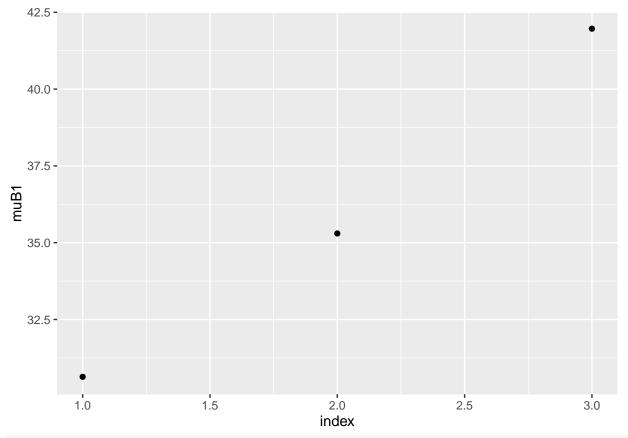
```
lm3.b<-lm(strength~A+as.factor(B),data=data3)</pre>
lm3.b
##
## Call:
## lm(formula = strength ~ A + as.factor(B), data = data3)
## Coefficients:
##
     (Intercept)
                               A as.factor(B)2
                                        -12.589
##
          12.044
                          5.667
data3.plot$muB1<-c(12.044+5.667,12.044+5.667*2,12.044+5.667*3)
data3.plot$muB2<-c(12.044+5.667,12.044+5.667*2,12.044+5.667*3)-12.589
ggplot(data3.plot,aes(x=index,y=muB1))+geom_point()
```



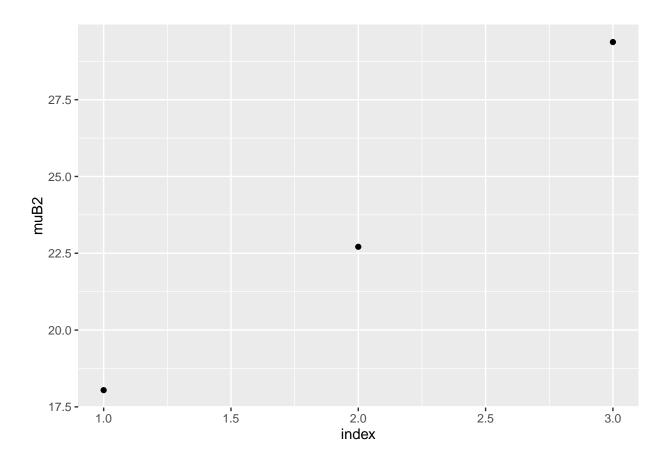
ggplot(data3.plot,aes(x=index,y=muB2))+geom_point()



```
c)
lm3.c<-lm(strength~as.factor(A)+B,data = data3)</pre>
lm3.c
##
## Call:
## lm(formula = strength ~ as.factor(A) + B, data = data3)
## Coefficients:
##
     (Intercept) as.factor(A)2 as.factor(A)3
                                                             В
          30.633
                          4.667
                                         11.333
                                                       -12.589
##
data3.plot$muB1<-c(30.633,30.633+4.667,30.633+11.333)
data3.plot$muB2<-c(30.633,30.633+4.667,30.633+11.333)-12.589
ggplot(data3.plot,aes(x=index,y=muB1))+geom_point()
```



ggplot(data3.plot,aes(x=index,y=muB2))+geom_point()



8.6

```
data8.6<-read.table("CHOSPRO6.txt")
colnames(data8.6)<-c("Y","X")</pre>
```